



Asymptotic stability and stationary states for stochastic systems: a pathwise approach.

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Mathematics of random systems: Analysis, modelling and algorithms

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Stochastic stability: moment approach

- Asymptotic stability for deterministic systems: Lyapunov (1892) $\epsilon - \delta$ language.
- Stochastic stability for Markov systems: Khasminskii (1979), Skorohod (1987), ...
- Exponential stability: Molchanov, Arnold, Pardoux, Baxendale, Mao... and many others in control theory.

$$dy_t = [Ay_t + f(y_t)]dt + g(y_t)dB_t, \quad x(0) = x_0 \in \mathbb{R}^d, \quad (1.1)$$

$A \in \mathbb{R}^{d \times d}$ neg. def. λ_A , $f: \mathbb{R}^d \rightarrow \mathbb{R}^d$, $g: \mathbb{R}^d \rightarrow \mathbb{R}^{d \times m}$ glob. Lischitz continuous w.r.t. C_f, C_g .

Applying Itô's formula

$$d\|y_t\|^2 = \left(2\langle y_t, Ay_t \rangle + 2\langle y_t, f(y_t) \rangle + \|g(y_t)\|^2 \right) dt + 2\langle y_t, g(y_t)dB_t \rangle \quad (1.2)$$

$$\leq \left(-2\lambda_A + 2C_f + C_g^2 \right) \|y_t\|^2 + \|g(y_t)\|^2 dt + 2\langle y_t, g(y_t)dB_t \rangle. \quad (1.3)$$

Applying the expectation, $\mathbb{E}\langle y_t, g(y_t)dB_t \rangle = 0$, thus

$$d\mathbb{E}\|y_t\|^2 \leq (-2\lambda_A + 2C_f + C_g^2)\mathbb{E}\|y_t\|^2 dt.$$

Applying continuous Gronwall lemma

$$\mathbb{E}\|y_t\|^2 \leq \mathbb{E}\|y_0\|^2 e^{-2(\lambda_A - C_f - \frac{1}{2}C_g^2)t}.$$

$\Rightarrow E\|y_t\|$ is exponentially decaying to zero given $\lambda_A - C_f > \frac{1}{2}C_g^2$, which follows that $\|y_t\|$ converges exponentially and almost surely to zero due to Borel-Catelli lemma.

Stochastic stability: pathwise approach

Pathwise approach

$$y_t = \Phi(t)y_0 + \int_0^t \Phi(t-s)f(y_s)ds + \int_0^t \Phi(t-s)g(y_s)dB_s$$

where $\Phi(t) = e^{At}$ is the semigroup generated by A . In addition, $\|\Phi(t)\| \leq C_A e^{-\lambda_A t}$, $\forall t \geq 0$. First, for any $t \in [n, n+1)$, it follows from (1.1) and the global Lipschitz continuity of f that

$$\begin{aligned} \|y_t\| &\leq \|\Phi(t)y_0\| + \int_0^t \|\Phi(t-s)f(y_s)\| ds + \left\| \int_0^t \Phi(t-s)g(y_s)dB_s \right\| \\ &\leq C_A e^{-\lambda_A t} \|y_0\| + \int_0^t C_A e^{-\lambda_A(t-s)} (C_f \|y_s\| + \|f(0)\|) ds + \left\| \int_0^t \Phi(t-s)g(y_s)dB_s \right\| \\ &\leq C_A e^{-\lambda_A t} \|y_0\| + \frac{C_A}{\lambda_A} \|f(0)\| (1 - e^{-\lambda_A t}) + C_A C_f \int_0^t e^{-\lambda_A(t-s)} \|y_s\| ds + \beta_t, \end{aligned}$$

where $\beta_t := \left\| \int_0^t \Phi(t-s)g(y_s)dB_s \right\|$. Multiplying both sides of the above inequality with $e^{\lambda_A t}$ yields

$$\|y_t\| e^{\lambda_A t} \leq C_A \|y_0\| + \frac{C_A}{\lambda_A} \|f(0)\| (e^{\lambda_A t} - 1) + \beta_t e^{\lambda_A t} + C_A C_f \int_0^t e^{\lambda_A s} \|y_s\| ds.$$

Stochastic stability: pathwise approach

By applying the continuous Gronwall Lemma, we obtain

$$\begin{aligned} \|y_t\| e^{\lambda_A t} &\leq C_A \|y_0\| + \frac{C_A}{\lambda_A} \|f(0)\| (e^{\lambda_A t} - 1) + \beta_t e^{\lambda_A t} \\ &\quad + \int_0^t C_A C_f e^{C_A C_f (t-s)} \left[C_A \|y_0\| + \frac{C_A}{\lambda_A} \|f(0)\| (e^{\lambda_A s} - 1) + \beta_s e^{\lambda_A s} \right] ds. \end{aligned}$$

Once again, multiplying both sides of the above inequality with $e^{-C_A C_f t}$ and assigning $\lambda := \lambda_A - C_A C_f$ yields

$$\begin{aligned} \|y_t\| e^{\lambda t} &\leq C_A \|y_0\| e^{-C_A C_f t} + \frac{C_A}{\lambda} \|f(0)\| (e^{\lambda t} - e^{-C_A C_f t}) + \beta_t e^{\lambda t} \\ &\quad + \int_0^t C_A C_f e^{-C_A C_f s} \left[C_A \|y_0\| + \frac{C_A}{\lambda} \|f(0)\| (e^{\lambda s} - 1) + \beta_s e^{\lambda s} \right] ds \\ &\leq C_A \|y_0\| + \frac{C_A}{\lambda} \|f(0)\| (e^{\lambda t} - 1) + \beta_t e^{\lambda t} + \int_0^t C_A C_f \beta_s e^{\lambda s} ds. \end{aligned} \quad (1.4)$$

Next, due to **isometry of Itô integral**

$$\mathbb{E} \beta_t^2 = \mathbb{E} \left\| \int_0^t \Phi(t-s) g(y_s) dB_s \right\|^2 = (C) \int_0^t \mathbb{E} \|\Phi(t-s) g(y_s)\|^2 ds \leq C_A^2 C_g^2 \int_0^t e^{-2\lambda_A(t-s)} \mathbb{E} \|y_s\|^2 ds.$$

⇒ **isometry assumption can be relaxed upto a constant C!**

Stochastic stability: pathwise approach

Then by applying the inequality

$$(a + b + c)^2 \leq (1 + \epsilon)a^2 + (1 + \frac{1}{\epsilon})(1 + \bar{\epsilon})b^2 + (1 + \frac{1}{\epsilon})(1 + \frac{1}{\bar{\epsilon}})c^2, \quad \forall a, b, c, \epsilon, \bar{\epsilon} > 0$$

and

$$\left(\int_0^t \|\beta_s\| e^{-\lambda(t-s)} ds \right)^2 \leq \left(\int_0^t e^{-2\varrho(t-s)} ds \right) \left(\int_0^t \|\beta_s\|^2 e^{-2(\lambda-\varrho)(t-s)} ds \right)$$

to (1.4), we obtain for any $0 < \varrho < \lambda$ and any $\epsilon, \bar{\epsilon} > 0$ the estimate

$$\begin{aligned} \mathbb{E}\|y_t\|^2 &\leq (1 + \frac{1}{\epsilon})(1 + \frac{1}{\bar{\epsilon}})C_A^2 e^{-2(\lambda-\varrho)t} \mathbb{E}\|y_0\|^2 + (1 + \epsilon)C_A^2 C_g^2 \int_0^t e^{-2(\lambda-\varrho)(t-s)} \mathbb{E}\|y_s\|^2 ds \\ &\quad + (1 + \frac{1}{\epsilon})(1 + \bar{\epsilon})C_A^2 C_g^2 \frac{L_f^2}{2\varrho} \int_0^t \int_0^s e^{-2(\lambda-\varrho)(t-u)} \mathbb{E}\|y_u\|^2 du ds. \end{aligned} \quad (1.5)$$

Multiplying both sides of (1.5) by $e^{2(\lambda-\varrho)t}$ and assigning $\gamma_t := e^{2(\lambda-\varrho)t} \mathbb{E}\|y_t\|^2$, we derive an inequality

$$\gamma_t \leq (1 + \frac{1}{\epsilon})(1 + \frac{1}{\bar{\epsilon}})C_A^2 \gamma_0 + (1 + \epsilon)C_A^2 C_g^2 \int_0^t \gamma_s ds + (1 + \frac{1}{\epsilon})(1 + \bar{\epsilon})C_A^2 C_g^2 \frac{L_f^2}{2\varrho} \int_0^t \int_0^s \gamma_u du ds.$$

Stochastic stability: pathwise approach

Lemma (Gronwall lemma for iterated integrals)

Assume that $T, a, b, c > 0$ and $\gamma_t > 0$ satisfies

$$\gamma_t \leq a\gamma_0 + b \int_0^t \gamma_s ds + c \int_0^t \int_0^s \gamma_u du ds, \quad \forall t \in [0, T]. \quad (1.6)$$

Then $\gamma_t \leq a\gamma_0 e^{\nu_2 t}$ for all $t \in [0, T]$, where $\nu_2 > 0 > \nu_1$ are two roots of the quadratic $\nu^2 - b\nu - c = 0$.

It then follows from the **iterated Gronwall lemma** that $\gamma_t < (1 + \frac{1}{\epsilon})(1 + \frac{1}{\bar{\epsilon}})C_A^2\gamma_0 e^{\nu_2 t}$, where $\nu_2 = \nu_2(\varrho, \epsilon, \bar{\epsilon})$ is the positive root of the quadratic equation

$$\nu^2 - (1 + \epsilon)C_A^2C_g^2\nu - (1 + \frac{1}{\bar{\epsilon}})(1 + \bar{\epsilon})C_A^2C_g^2\frac{L_f^2}{2\varrho} = 0. \quad (1.7)$$

Hence $\mathbb{E}\|y_t\|^2 \leq (1 + \frac{1}{\epsilon})(1 + \frac{1}{\bar{\epsilon}})C_A^2\mathbb{E}\|y_0\|^2 e^{-2(\lambda - \varrho - \frac{\nu_2}{2})t}$, which derives the sufficient condition for the exponential stability as follows

$$\lambda = \lambda_A - C_A C_f > \inf_{\substack{0 < \varrho < \lambda, \\ \epsilon > 0}} \left\{ \varrho + \frac{1}{4}(1 + \epsilon)C_A^2C_g^2 + \frac{1}{4}C_A^2C_g \left[(1 + \epsilon)^2 C_g^2 + \frac{2C_f^2}{\varrho} \left(1 + \frac{1}{\bar{\epsilon}}\right) \right]^{\frac{1}{2}} \right\}. \quad (1.8)$$

\Rightarrow compare with $\lambda = \lambda_A - C_A C_f > \frac{1}{2}C_g^2$: **still good enough with relaxed Itô isometry!**

Stochastic stability: pathwise approach

Question: How to estimate $\left\| \int_0^t \Phi(t-s)g(y_s)dB_s \right\|$ in general? Answer: rough path theory!

(i) The following estimate holds: for any $0 \leq a < b \leq c$

$$\left\| \int_a^b \Phi(c-s)g(y_s)dx_s \right\| \leq e^{-\lambda_A(c-b)} \kappa(\mathbf{x}, [a, b]) \left(\frac{\|g(0)\|}{C_g} + \|y_a\| + \|y, R^y\|_{\rho\text{-var}, [a, b]} \right), \quad (1.9)$$

where

$$\kappa(\mathbf{x}, [a, b]) := 2C_p C_A [1 + \|A\|(b-a)] \left\{ C_g^2 \|\mathbf{x}\|_{\rho\text{-var}, [a, b]}^2 \vee C_g \|\mathbf{x}\|_{\rho\text{-var}, [a, b]} \right\}. \quad (1.10)$$

(ii) $z_t = \bar{y}_t - y_t$. Then for any $0 \leq a < b \leq c$

$$\left\| \int_a^b \Phi(c-s)[g(\bar{y}_s) - g(y_s)]dx_s \right\| \leq e^{-\lambda_A(c-b)} \kappa(\mathbf{x}, [a, b]) \Lambda(\mathbf{x}, [a, b]) \left(\|z_a\| + \|z, R^z\|_{\rho\text{-var}, [a, b]} \right), \quad (1.11)$$

for some function Λ . $\|z, R^z\|_{\rho\text{-var}, [a, b]}$ can be estimated by z_a .

$$\begin{aligned} \text{(iii)} \quad \|y_t\| e^{\lambda t} &\leq C_A \|y_0\| + \frac{C_A}{\lambda_A - C_A C_f} \|f(0)\| (e^{\lambda t} - 1) \\ &\quad + e^{\lambda A} \sum_{k=0}^n e^{\lambda k} \kappa(\mathbf{x}, \Delta_k) \left[\frac{\|g(0)\|}{C_g} + \|y_k\| + \|y, R^y\|_{\rho\text{-var}, \Delta_k} \right], \quad \forall t \in \Delta_n \end{aligned} \quad (1.12)$$

where $\Delta_k := [k, k+1]$, $\lambda := \lambda_A - C_A C_f$, and $\|y, R^y\|_{\rho\text{-var}, \Delta_k}$ depends linearly on $y_k \Rightarrow$ discrete Gronwall lemma to obtain the absorbing set. (details in next slide)

Stochastic stability: pathwise approach

Assign $a := C_A \|y_0\|$, $u_k := \|y_k\| e^{\lambda k}$, $k \geq 0$, one obtains

$$u_n \leq a + \sum_{k=0}^{n-1} G(\mathbf{x}, \Delta_k) u_k + \sum_{k=0}^{n-1} e^{\lambda k} H(\mathbf{x}, \Delta_k). \quad (1.13)$$

Discrete Gronwall Lemma: Let a be a non-negative constant and u_n, α_n, β_n be non-negative sequences satisfying

$$u_n \leq a + \sum_{k=0}^{n-1} \alpha_k u_k + \sum_{k=0}^{n-1} \beta_k, \quad \forall n \geq 1$$

then

$$u_n \leq \max\{a, u_0\} \prod_{k=0}^{n-1} (1 + \alpha_k) + \sum_{k=0}^{n-1} \beta_k \prod_{j=k+1}^{n-1} (1 + \alpha_j), \quad \forall n \geq 1. \quad (1.14)$$

By assigning ω with $\theta_{-n}\omega$ and using the shift property

$$\|y_n(\mathbf{x}(\theta_{-n}\omega), y_0(\theta_{-n}\omega))\| \leq C_A \|y_0(\theta_{-n}\omega)\| e^{-\lambda n} \prod_{k=0}^{n-1} \left[1 + G(\mathbf{x}(\theta_{-k}\omega), [-1, 0]) \right] + b(\omega), \quad (1.15)$$

where

$$b(\omega) := \sum_{k=1}^{\infty} e^{-\lambda k} H(\mathbf{x}(\theta_{-k}\omega), [-1, 0]) \prod_{j=0}^{k-1} \left[1 + G(\mathbf{x}(\theta_{-j}\omega), [-1, 0]) \right]. \quad (1.16)$$

Norms and spaces

- For $p \geq 1$, $C^{p\text{-var}}([a, b], H)$: the Banach space with finite norm

$$\|u\|_{p,[a,b]} = \|u_a\| + \|u\|_{p,[a,b]} = \|u_a\| + \|u\|_{p,[a,b]} = \left(\sup_{\Pi(a,b)} \sum_{[s,t] \in \Pi} |u_{s,t}|^p \right)^{1/p} < \infty.$$

- For $\beta \in (0, 1)$, $C^{\beta\text{-Hol}}([a, b], H)$ the Banach space equipped with norm

$$\|u\|_{\infty,\beta,[a,b]} := \|u\|_{\infty,[a,b]} + \|u\|_{\beta,[a,b]} = \|u\|_{\infty,[a,b]} + \sup_{a \leq s < t \leq b} \frac{\|u(t) - u(s)\|}{(t-s)^\beta}.$$

- For $\alpha, \beta \in (0, 1)$, $C^{\alpha,\beta}([a_+, b], H)$, $C^{\alpha,\beta}([a, b_-], H)$ the Banach spaces equipped with norms

$$\|u\|_{\infty,\alpha,\beta,[a_+,b]} := \|u\|_{\infty,[a,b]} + \sup_{a < s < t \leq b} |s-a|^\alpha \frac{\|u(t) - u(s)\|}{(t-s)^\beta}$$

$$\|u\|_{\infty,\alpha,\beta,[a,b_-]} := \|u\|_{\infty,[a,b]} + \sup_{a \leq s < t < b} |b-t|^\alpha \frac{\|u(t) - u(s)\|}{(t-s)^\beta}.$$

- One can prove that $\|y\|_{\alpha,\beta,[a_+,b]} = \sup_{a < s < b} (s-a)^\alpha \|y\|_{\beta,[s,b]}$, thus

$$C^\beta([a, b]) \subset C^{\beta,\beta}([a_+, b]) \cap C^{\beta,\beta}([a, b_-]) \text{ with } \|u\|_{\beta,\beta,[a_+,b]} \leq |b-a|^\beta \|u\|_{\beta,[a,b]}.$$

Rough paths

In the simplest form for $\beta \in (\frac{1}{3}, \frac{1}{2})$, a couple $\mathbf{x} = (x, \mathbb{X})$, with $x \in C^\beta([a, b], H)$ and $\mathbb{X} \in C_2^{2\beta}([a, b]^2, H \otimes H)$ is called a *rough path* if they satisfy Chen's relation

$$\mathbb{X}_{s,t} - \mathbb{X}_{s,u} - \mathbb{X}_{u,t} = x_{s,u} \otimes x_{u,t}, \quad \forall a \leq s \leq u \leq t \leq b. \quad (2.1)$$

We write $\mathbb{X}_{s,t} = (x \otimes x)_{s,t} =: \int_s^t x_{s,u} \otimes dx_u$. In a general form, $(y \otimes x)_{s,t} =: \int_s^t y_{s,u} \otimes dx_u$ satisfies Chen's relation

$$(y \otimes x)_{s,t} - (y \otimes x)_{s,u} - (y \otimes x)_{u,t} = y_{s,u} \otimes x_{u,t}.$$

Denote by $C^\beta([a, b]) \subset C^\beta \oplus C^{2\beta}$ the set of all rough paths in $[a, b]$, then C^β is a closed set but not a linear space, equipped with the rough path semi-norm

$$\|\mathbf{x}\|_{\beta, [a, b]} := \|x\|_{\beta, [a, b]} + \|\mathbb{X}\|_{2\beta, [a, b]}^{\frac{1}{2}} < \infty. \quad (2.2)$$

Examples in 1D

- $X_t = \int_0^t a_s dB_s \Rightarrow \mathbb{X}_{s,t} := \int_s^t X_{s,u} dX_u = \frac{1}{2} X_{s,t}^2 - \frac{1}{2} \int_s^t a_u^2 du$. Integral in the Itô sense.
- $X_t = B^H \Rightarrow \mathbb{X}_{s,t} := \int_s^t B_{s,u}^H \delta B_u^H = \frac{1}{2} (B_{s,t}^H)^2 - \frac{1}{2} (t^{2H} - s^{2H})$. Integral in the Skorohod-Wick-Itô sense.
- $[x, 2]_{s,t} := x_{s,t}^2 - 2\mathbb{X}_{s,t} \Rightarrow [x, 2]_{s,t} = [x, 2]_{s,u} + [x, 2]_{u,t} \Rightarrow [x, 2]_{0,\cdot} \in C^{2\nu}$ is a Hölder continuous path.

$\beta > \frac{1}{2}$: Young integrals

- Consider $y \in C^\beta([a, b], V \otimes H)$ and $x \in C^\beta([a, b], H)$, there exists a Young integral $\int_a^b y dx$ defined by

$$\int_a^b y dx = \lim_{|\Pi| \rightarrow 0} \sum_{[s,t] \in \Pi} y_s x_{s,t},$$

- Young-Loeve (Y-L) estimate

$$\left\| \int_s^t y_u dx_u - y_s x_{s,t} \right\| \leq K_\beta |t - s|^{2\beta} \|y\|_{\beta, [a,b]} \|x\|_{\beta, [a,b]}.$$

In fact, we can prove using fractional calculus that

$$\left\| \int_s^t y_u dx_u - y_s x_{s,t} \right\| \leq K_\beta \left(\|y_s \cdot\|_{\infty, [s,t]} + \|y\|_{\beta, \beta, [s_+, t]} \right) \left(\|x_{\cdot, t}\|_{\infty, [s,t]} + \|x\|_{\beta, \beta, [s, t_-]} \right).$$

$\beta \in (\frac{1}{3}, \frac{1}{2})$: Rough integrals

- A path $y \in C^\beta([a, b], V \otimes H)$ is then called to be *controlled* by $x \in C^\beta([a, b], H)$ if there exists a tube (y', R^y) with $y' \in C^\beta([a, b], V \otimes H \otimes H)$, $R^y \in C^{2\beta}(\Delta^2([a, b]), V \otimes H)$ such that

$$y_{s,t} = y'_s x_{s,t} + R^y_{s,t}, \quad \forall \min l \leq s \leq t \leq \max l.$$

y' is called **Gubinelli derivative** of y , which is uniquely defined as long as $x \in C^\beta \setminus C^{2\beta}$ is truly rough (works for fBm B^H , $H \in (\frac{1}{3}, \frac{1}{2}]$ FH[Chapter 6]).

- Denote $F_{s,t} := y_s x_{s,t} + y'_s \mathbb{X}_{s,t}$. Then

$$\begin{aligned} \delta F_{s,u,t} &:= F_{s,t} - F_{s,u} - F_{u,t} = -y'_{s,u} \mathbb{X}_{u,t} - R^y_{s,u} x_{u,t} \\ \Rightarrow \|\delta F_{s,u,t}\| &\leq |t-s|^{3\beta} \left(\|x\|_{\beta,[s,t]} \|R^y\|_{2\beta,\Delta^2[s,t]} + \|y'\|_{\beta,[s,t]} \|\mathbb{X}\|_{2\beta,\Delta^2[s,t]} \right) \end{aligned}$$

Thanks to the sewing lemma, the rough integral $\int_s^t y_u dx_u$ can be defined as

$$\int_s^t y_u dx_u := \lim_{|\Pi| \rightarrow 0} \sum_{[u,v] \in \Pi} [y_u x_{u,v} + y'_u \mathbb{X}_{u,v}]$$

Moreover, one gets Y-L typed estimate:

$$\left\| \int_s^t y_u dx_u - y_s x_{s,t} - y'_s \mathbb{X}_{s,t} \right\| \leq C_\beta |t-s|^{3\beta} \left(\|x\|_{\beta,[s,t]} \|R^y\|_{2\beta,\Delta^2[s,t]} + \|y'\|_{\beta,[s,t]} \|\mathbb{X}\|_{2\beta,\Delta^2[s,t]} \right).$$

Existence and uniqueness theorem: RDEs

- Finite dimension: RDE $dy_t = f(y_t)dt + g(y_t)dx_t$ is understood in the integral form

$$y_t = y_0 + \int_0^t f(y_s)ds + \int_0^t g(y_s)dx_s. \quad (2.3)$$

- Riedel and Scheutzow (2017): f is one sided Lipschitz and linear growth in perpendicular direction. $g \in C_b^3$ as usual. Solution of the form (y, y') with $y, y' \in C^\beta$, $y' = g(y)$, which is solved by Doss-Sussmann technique.
- Infinite dimension: consider rough evolution equation

$$dy_t = [Ay_t + f(y_t)]dt + g(y_t)dx_t, \quad (2.4)$$

with the solution understood in the mild form

$$y_t = S(t)y_0 + \int_0^t S(t-s)f(y_s)ds + \int_0^t S(t-s)g(y_s)dx_s. \quad (2.5)$$

Key task: define $\int_0^t S(t-s)g(y_s)dx_s$ for $S(\cdot) \in C^{\beta, \beta}$.

- Garrido-Atienza & Lu & Schmalfuss (2015): define rough integral using fractional calculus and solution definition by Hu & Nualart (2009).
- Hesse & Neamtu (2019): define rough integrals for controlled rough path $y \in C^{\beta, \beta}$ and $x \in C^\beta$ using a modified sewing lemma.
- Our idea: **A hybrid form**, i.e. define rough integral for controlled rough path using fractional calculus. Advantage: **simpler scheme with explicit Y-L typed estimate, and solution (y, y')** .

Fixed point argument

- Back to Gubinelli: solve the equation on some small time interval $[0, \tau]$ for $(y, y') \in \mathcal{M}_x^{\beta, \beta}(y_0, g(y_0))$ such that $\| (y, y') \|_{\beta, \beta, [0_+, \tau]} \leq 1$ and

$$y_t = S(t)y_0 + \int_0^t S(t-s)f(y_s)ds + \int_0^t S(t-s)g(y_s)dx_s,$$

$$y'_t = g(y_t).$$

Many ways to choose seminorm, but the simplest one (works well for $C_g < 1$) is

$$\| (y, y') \|_{\beta, \beta, [0_+, \tau]} := \| y' \|_{\beta, \beta, [0_+, \tau]} + \| y'_0 \cdot \|_{\infty, [0_+, \tau]} + \| R^y \|_{\beta, 2\beta, [0_+, \tau]} + \| R_{0, \cdot}^y \|_{\infty, [0_+, \tau]} \leq 1.$$

For general C_g , better use the exponential weighted seminorm $\| (y, y') \|_{\beta, \beta, \rho, [0_+, \tau]}$.

- Examples in 1D: $dy = \mu y dt + \sigma y dx_t \Rightarrow y_t = y_s \exp\{\mu(t-s) - \frac{\sigma^2}{2}[X, 2]_{s,t} + \sigma X_{s,t}\}$ where $[X, 2]_{s,t} := x_{s,t}^2 - 2\mathbb{X}_{s,t}$.

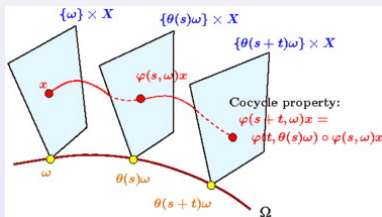
Random dynamical systems

Definition

A (continuous) random dynamical system is an NDS (θ, φ) which in addition has the following properties:

- (i), The driving system θ is an ergodic dynamical system $(\Omega, \mathcal{F}, \mathbb{P}, (\theta_t)_{t \in T})$, i.e., the base $(\Omega, \mathcal{F}, \mathbb{P})$ is a probability space and $(t, \omega) \mapsto \theta_t \omega$ is a measurable flow which is ergodic under \mathbb{P} .
- ii, The cocycle $(t, \omega, x) \mapsto \varphi(t, \omega)x$ is measurable w.r.t $\mathcal{B} \otimes \mathcal{F} \otimes \mathcal{B}(X)$ and $\mathcal{B}(X)$, respectively, where $\mathcal{B}(X)$ is the Borel σ -algebra of X , such that the family $\varphi(t, \omega, \cdot) = \varphi(t, \omega) : X \rightarrow X$ of self-mappings of X satisfies the *cocycle property*

$$\varphi(0, \omega) = id_X, \quad \varphi(t + s, \omega) = \varphi(t, \theta_s \omega) \circ \varphi(s, \omega), \quad \forall t, s \in T, \quad \forall \omega \in \Omega. \quad (3.1)$$



Generation of RDS from RDE

- Finite dimension - Bailleul & Riedel & Scheutzow (2018)
- Infinite dimensional case: Garrido-Atienza & Lu & Schmalfuss (2010,2016), Hesse & Neamtu (2020)...

$T_1^2(\mathbb{R}^m) = \mathbf{1} \oplus \mathbb{R}^m \oplus (\mathbb{R}^m \otimes \mathbb{R}^m)$, is the set with the tensor product

$$(1, g^1, g^2) \otimes (1, h^1, h^2) = (1, g^1 + h^1, g^1 \otimes h^1 + g^2 + h^2), \quad \forall \mathbf{g} = (1, g^1, g^2), \mathbf{h} = (1, h^1, h^2) \in T_1^2(\mathbb{R}^m).$$

Then it can be shown that $(T_1^2(\mathbb{R}^m), \otimes)$ is a topological group with unit element $\mathbf{1} = (1, 0, 0)$.

For $\beta \in (\frac{1}{p}, \nu)$, denote by $\mathcal{C}^{0, \beta\text{-var}}([a, b], T_1^2(\mathbb{R}^m))$ the closure of $\mathcal{C}^\infty([a, b], T_1^2(\mathbb{R}^m))$ in

$\mathcal{C}^{\beta\text{-var}}([a, b], T_1^2(\mathbb{R}^m))$, and by $\mathcal{C}_0^{0, \beta\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m))$ the space of all $x : \mathbb{R} \rightarrow \mathbb{R}^m$ such that

$x|_I \in \mathcal{C}^{0, \beta\text{-var}}(I, T_1^2(\mathbb{R}^m))$ for each compact interval $I \subset \mathbb{R}$ containing 0. Then $\mathcal{C}_0^{0, \beta\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m))$ is equipped with the compact open topology given by the β -variation norm, i.e the topology generated by the metric:

$$d_\beta(\mathbf{x}_1, \mathbf{x}_2) := \sum_{k \geq 1} \frac{1}{2^k} (\|\mathbf{x}_1 - \mathbf{x}_2\|_{\beta\text{-var}, [-k, k]} \wedge 1).$$

As a result, it is separable and thus a Polish space.

Let us consider a stochastic process $\bar{\mathbf{X}}$ defined on a probability space $(\bar{\Omega}, \bar{\mathcal{F}}, \bar{\mathbb{P}})$ with realizations in $(\mathcal{C}_0^{0, \beta\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m)), \mathcal{F})$. Assume further that $\bar{\mathbf{X}}$ has stationary increments.

Generation of RDS: rough equations

Assign

$$\Omega := C_0^{0,p\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m))$$

and equip with the Borel σ -algebra \mathcal{F} and let \mathbb{P} be the law of $\bar{\mathbf{X}}$. Denote by θ the *Wiener-type shift*

$$(\theta_t \omega)_\cdot = \omega_t^{-1} \otimes \omega_{t+\cdot}, \forall t \in \mathbb{R}, \omega \in C_0^{0,p\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m)), \quad (3.2)$$

and define the so-called *diagonal process* $\mathbf{X} : \mathbb{R} \times \Omega \rightarrow T_1^2(\mathbb{R}^m)$, $\mathbf{X}_t(\omega) = \omega_t$ for all $t \in \mathbb{R}$, $\omega \in \Omega$.

Due to the stationarity of $\bar{\mathbf{X}}$, it can be proved that θ is invariant under \mathbb{P} , then forming a continuous (and thus measurable) dynamical system on $(\Omega, \mathcal{F}, \mathbb{P})$ [BRSch17 -Theorem 5]. Moreover, \mathbf{X} forms a p -rough path cocycle, namely, $\mathbf{X}_\cdot(\omega) \in C_0^{0,p\text{-var}}(\mathbb{R}, T_1^2(\mathbb{R}^m))$ for every $\omega \in \Omega$, which satisfies the *cocycle relation*:

$$\mathbf{X}_{t+s}(\omega) = \mathbf{X}_s(\omega) \otimes \mathbf{X}_t(\theta_s \omega), \forall \omega \in \Omega, t, s \in \mathbb{R},$$

in the sense that $\mathbf{X}_{s,s+t} = \mathbf{X}_t(\theta_s \omega)$ with the increment notation $\mathbf{X}_{s,s+t} := \mathbf{X}_s^{-1} \otimes \mathbf{X}_{s+t}$. It is important to note that the two-parameter flow property

$$\mathbf{X}_{s,u} \otimes \mathbf{X}_{u,t} = \mathbf{X}_{s,t}, \forall s, t \in \mathbb{R}$$

is equivalent to the fact that $\mathbf{X}_t(\omega) = (1, x_t(\omega), \mathbb{X}_{0,t}(\omega))$, where $x_\cdot(\omega) : \mathbb{R} \rightarrow \mathbb{R}^m$ and $\mathbb{X}_\cdot(\omega) : I \times I \rightarrow \mathbb{R}^m \otimes \mathbb{R}^m$ are random functions satisfying Chen's relation (2.1).

Generation of RDS: rough equations

In particular, due to the fact that $\|\mathbf{X} \cdot (\theta_h \omega)\|_{p\text{-var}, [s, t]} = \|\mathbf{X} \cdot (\omega)\|_{p\text{-var}, [s+h, t+h]}$, it follows from Birkhoff ergodic theorem that

$$\Gamma(\mathbf{x}, p) := \limsup_{n \rightarrow \infty} \left(\frac{1}{n} \sum_{k=1}^n \|\theta_{-k} \mathbf{x}\|_{p\text{-var}, [-1, 1]}^p \right)^{\frac{1}{p}} = \left(E \|\mathbf{X} \cdot (\cdot)\|_{p\text{-var}, [-1, 1]}^p \right)^{\frac{1}{p}} = \Gamma(p) \quad (3.3)$$

for almost all realizations \mathbf{x}_t of the form $\mathbf{X}_t(\omega)$. We assume additionally that $(\Omega, \mathcal{F}, \mathbb{P}, \theta)$ is ergodic.

Theorem

Let $(\Omega, \mathcal{F}, \mathbb{P}, \theta)$ be a measurable metric dynamical system and let $\mathbf{X} : \mathbb{R} \times \Omega \rightarrow T_1^2(\mathbb{R}^m)$ be a p -rough cocycle for some $2 \leq p < 3$. Then there exists a unique continuous random dynamical system φ over $(\Omega, \mathcal{F}, \mathbb{P}, \theta)$ which solves the rough differential equation

$$dy_t = f(y_t)dt + g(y_t)d\mathbf{X}_t(\omega), t \geq 0. \quad (3.4)$$

Random attractors

Universe: \mathcal{D} a family of random sets $\hat{M} = \{M(\omega)\}_{\omega \in \Omega}$ which is closed w.r.t. inclusions (i.e. if $\hat{D}_1 \in \mathcal{D}$ and $\hat{D}_2 \subset \hat{D}_1$ then $\hat{D}_2 \in \mathcal{D}$).

Definition (Crauel & Flandoli-1994)

An invariant random compact set $\hat{A} \in \mathcal{D}$ is called

(i), a pullback random attractor in \mathcal{D} , if \hat{A} attracts any closed random set $\hat{D} \in \mathcal{D}$ in the pullback sense, i.e.

$$\lim_{t \rightarrow \infty} d(\varphi(t, \theta_{-t}\omega)D(\theta_{-t}\omega) | A(\omega)) = 0; \quad (4.1)$$

(ii), a forward random attractor in \mathcal{D} , if \hat{A} attracts any closed random set $\hat{D} \in \mathcal{D}$ in the forward sense, i.e.

$$\lim_{t \rightarrow \infty} d(\varphi(t, \omega)D(\omega) | A(\theta_t\omega)) = 0; \quad (4.2)$$

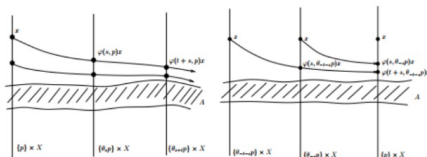


Figure 2.1: Forward attraction versus pullback attraction.

Existence of random attractor

Pullback absorbing set: $\hat{B} \in \mathcal{D}$ in a universe \mathcal{D} such that \hat{B} absorbs all sets in \mathcal{D} , i.e. for any $\hat{D} \in \mathcal{D}$, there exists a time $t_0 = t_0(\omega, \hat{D})$ such that

$$\varphi(t, \theta_{-t}\omega)D(\theta_{-t}\omega) \subset B(\omega), \text{ for all } t \geq t_0.$$

Theorem (Crauel, Flandoli, Schenk-Hoppe 1998: Existence and Uniqueness of Random Attractor)

Given a universe \mathcal{D} , assume there exists a random compact pullback absorbing set $\hat{B} \in \mathcal{D}$ which is forward invariant. Then there exists a unique random pullback attractor (which is then a weak attractor) in \mathcal{D} , given by

$$A(\omega) = \bigcap_{t \geq 0} \varphi(t, \theta_{-t}\omega)B(\theta_{-t}\omega).$$

Application: prove that a system has an absorbing set.

Random attractors for RDE: finite dimension

We would like to investigate the RDE of the form

$$dy_t = f(y_t)dt + g(y_t)dx_t. \quad (4.3)$$

(H_f) f is strongly dissipative, i.e. there exists $D_1, D_2 > 0$ such that

$$\langle y_1 - y_2, f(y_1) - f(y_2) \rangle \leq D_1 - D_2 \|y_1 - y_2\|^2, \quad \forall y_1, y_2 \in \mathbb{R}^d; \quad (4.4)$$

in addition f is of linear growth in the perpendicular direction, i.e. there exists $C_f > 0$ such that

$$\left\| f(y_1) - f(y_2) - \frac{\langle f(y_1) - f(y_2), y_1 - y_2 \rangle (y_1 - y_2)}{\|y_1 - y_2\|^2} \right\| \leq C_f (1 + \|y_1 - y_2\|), \quad \forall y_1 \neq y_2; \quad (4.5)$$

(H_g) g belongs to $C_b^3(\mathbb{R}^d, (\mathbb{R}^m, \mathbb{R}^d))$ such that

$$C_g := \max \left\{ \|g\|_\infty, \|Dg\|_\infty, \|D_g^2\|_\infty, \|D_g^3\|_\infty \right\} < \infty; \quad (4.6)$$

(H_X) for a given $\nu \in (\frac{1}{3}, \frac{1}{2})$, x belongs to the space $C^\nu(\mathbb{R}, \mathbb{R}^m)$ of all continuous paths which is of finite ν -Hölder norm on any interval $[s, t]$. In particular, x is a realization of a stationary stochastic process $X_t(\omega)$, such that x can be lifted into a realized component $\mathbf{x} = (x, \mathbb{X})$ of a stationary stochastic process $(x \cdot (\omega), \mathbb{X} \cdot (\omega))$, such that the estimate

$$E \left(\|x_{s,t}\|^p + \|\mathbb{X}_{s,t}\|^q \right) \leq C_{T,\nu} |t - s|^{p\nu}, \quad \forall s, t \in [0, T] \quad (4.7)$$

holds for any $[0, T]$, with $p\nu \geq 1$, $q = \frac{p}{2}$ and some constant $C_{T,\nu}$.

Random attractors for RDE: finite dimension

$(\mathbf{H}_{\mathcal{A}})$ There exists a duration $r > 0$ and constants $D_3 > 0$ of the deterministic system such that, for any starting point $y_0 \notin \mathcal{A}$, there exists a point $\mu_0 = \mu_0(y_0) \in \mathcal{A}$ satisfying

$$\|\mu_r(y_0) - \mu_r(\mu_0)\| \leq e^{-D_3} \|y_0 - \mu_0\|. \quad (4.8)$$

Theorem (LHD-2020: finite dimension)

Assume that system (3.4) satisfies the assumptions (\mathbf{H}_f) , (\mathbf{H}_g) , (\mathbf{H}_X) , then there exists a random pullback attractor $\mathcal{A}(\omega)$ such that $|\mathcal{A}(\cdot)| \in \mathcal{P}$ for any $\rho \geq 1$. If in addition $(\mathbf{H}_{\mathcal{A}})$ and f is global Lipschitz continuous then the random attractor is upper semi-continuous with respect to the noise intensity in the sense that $\mathcal{A}(\omega) \rightarrow \mathcal{A}$ (w.r.t. the Hausdorff semi-distance) as $C_g \rightarrow 0$, both in the almost sure and in \mathcal{L}^p senses. Moreover, if f is strictly dissipative then $\mathcal{A}(\omega)$ is a singleton provided that C_g is sufficiently small.

Sketch of the proof: **Doss-Sussmann technique** to conjugate the RDE to a random differential equations. Under the assumptions (\mathbf{H}_f) , (\mathbf{H}_g) , (\mathbf{H}_X) , $(\mathbf{H}_{\mathcal{A}})$ and $f \in Lip$, the random attractor is upper semi-continuous, i.e.

$$\lim_{\bar{C}_g \rightarrow 0} d_H(\mathcal{A}(\omega) | \mathcal{A})^\rho = 0 \quad \text{a.s.} \quad \text{and} \quad \lim_{\bar{C}_g \rightarrow 0} \mathbb{E} d_H(\mathcal{A}(\cdot) | \mathcal{A})^\rho = 0, \quad \forall \rho \geq 1. \quad (4.9)$$

Random diffeomorphism

- (H_f) implies the existence of a global attractor \mathcal{A} of $\dot{\mu} = f(\mu)$. Example: $f(y) = \alpha y - y^3$.
- the solution $\phi \cdot (\mathbf{x}, \phi_a)$ of $d\phi_u = g(\phi_u)dx_u$, $u \in [a, b]$, $\phi_a \in \mathbb{R}^d$ is C^1 w.r.t. ϕ_a , and $\frac{\partial \phi}{\partial \phi_a}(\cdot, \mathbf{x}, \phi_a)$ is the solution of the linearized system

$$d\xi_u = Dg(\phi_u(\mathbf{x}, \phi_a))\xi_u dx_u, \quad u \in [a, b], \xi_a = Id, \quad (4.10)$$

where $Id \in \mathbb{R}^{d \times d}$ denotes the identity matrix. Moreover

$$\|\phi_b(\mathbf{x}, \phi_a) - \phi_a\| \leq \|\phi\|_{p\text{-var}, [a, b]} \leq 8C_p C_g \|\mathbf{x}\|_{p\text{-var}, [a, b]}; \quad (4.11)$$

$$\left\| \frac{\partial \phi}{\partial \phi_a}(t, \mathbf{x}, \phi_a) - Id \right\| \leq 16C_p C_g \|\mathbf{x}\|_{p\text{-var}, [a, b]}. \quad (4.12)$$

- Cass-Litterer-Lyons (2013): **Greedy times**

$$\tau_0 = \min I, \quad \tau_{i+1} := \inf \left\{ t > \tau_i : \|\mathbf{x}\|_{p\text{-var}, [\tau_i, t]} = \gamma \right\} \wedge \max I. \text{ Assign}$$

$$N(\gamma, \mathbf{x}, I) := \sup \{ j \in \mathcal{N} : \tau_j \leq \max I \}.$$

Lemma

For any $\lambda > 0$ small enough, there exist constants $\delta_\lambda, C_\lambda > 0$ such that for any solution μ_t of the ODE lying in the global attractor \mathcal{A} , the following estimates hold

$$\|y_t - \mu_t\| \leq \|y_0 - \mu_0\| e^{-\delta_\lambda t} + C_\lambda N\left(\frac{\lambda}{16C_p C_g}, \mathbf{x}, [0, t]\right). \quad (4.13)$$

Sketch of the proof

- Associate each solution $y_t(\mathbf{x}, y_0)$ with a solution $\mu_t(\mu_0)$ of the deterministic system $\dot{\mu} = \bar{f}(\mu)$ which starts at μ_0 . Consider the difference $y_t^* := y_t(y_0) - \mu_t(\mu_0)$ for $t \geq 0$. Similar to Hairer & Ohashi (2007), the key point is to prove that for any $\rho \geq 1$ there exists an $\eta \in (0, 1)$ and an integrable random variable $\xi_1(\omega) = \xi_1(C_g \|\mathbf{x}(\omega)\|_{\rho\text{-var}, [0,1]})$ such that

$$\|y_1^*\|^\rho \leq \eta \|y_0^*\|^\rho + \xi_1(\omega). \quad (4.14)$$

- Assign $\mu_t^* = \mu_t(y_0) - \mu_t(\mu_0)$ and $h_t := y_t^* - \mu_t^*$, then h satisfies

$$h_{0,t} = \int_0^t \left[f(h_u + \mu_u + \mu_u^*) - f(\mu_u + \mu_u^*) \right] du + \int_0^t g(h_u + \mu_u + \mu_u^*) dx_u.$$

Y-L estimate gives

$$\begin{aligned} \|h_{s,t}\| &\leq \int_s^t L_f \|h_u\| du + C_g \|x_{s,t}\| + C_g^2 \|\mathbb{X}_{s,t}\| + C_p \left\{ \|\mathbb{X}\|_{q\text{-var}, [s,t]^2} \|[g(y)]'\|_{\rho\text{-var}, [s,t]} \right. \\ &\quad \left. + \|x\|_{\rho\text{-var}, [s,t]} \|\|R^{g(y)}\|\|_{q\text{-var}, [s,t]^2} \right\}. \end{aligned} \quad (4.15)$$

Sketch of the proof

- One can then prove that

$$\begin{aligned} \left\| h, R^h \right\|_{p\text{-var}, [s, t]} &\leq \int_s^t 2L_f \left(\|h_s\| + \left\| h, R^h \right\|_{p\text{-var}, [s, u]} \right) du + \frac{1}{2} \left\| h, R^h \right\|_{p\text{-var}, [s, t]} \\ &\quad + \underbrace{4C_p C_g \|\mathbf{x}\|_{p\text{-var}, [s, t]} \left(1 + 4C_p \|\mu\|_{1\text{-var}, [s, t]} + 4C_p D(1 + \|y_0^*\|^\beta) \right)}_{=: L_1} \end{aligned} \quad (4.16)$$

whenever $4C_p C_g \|\mathbf{x}\|_{p\text{-var}, [s, t]} \leq \frac{1}{2}$, thus by the continuous Gronwall lemma,

$$\|h_s\| + \left\| h, R^h \right\|_{p\text{-var}, [s, t]} \leq (\|h_s\| + 2L_1) e^{4L_f(t-s)}$$

whenever $4C_p C_g \|\mathbf{x}\|_{p\text{-var}, [s, t]} \leq \frac{1}{2}$. Greedy time technique yields

$$\|h_r\| \leq \underbrace{e^{4L_f r} \left(1 + 4C_p r \|f\|_{\infty, \mathcal{A}} + 4C_p D \right) 8C_p C_g \|\mathbf{x}\|_{p\text{-var}, [0, r]} N\left(\frac{1}{8C_p C_g}, \mathbf{x}, [0, r]\right)}_{=: \xi_0(\mathbf{x})} (1 + \|y_0^*\|^\beta). \quad (4.17)$$

- From $(\mathbf{H}_{\mathcal{A}})$ one can choose μ_0 depending on y_0 such that $\|\mu_r^*\| \leq \|\mu_0^*\| e^{-D_2 r}$. Jensen's inequality for $\|y_r^*\|^\rho \leq (\|h_r\| + \|\mu_r^*\|)^\rho$ then derives (4.14).

Rough evolution equation

We would like to investigate the rough evolution equation

$$y_t = S(t)y_0 + \int_0^t S(t-u)f(y_u)du + \int_0^t S(t-u)g(y_u)dB_u^H, \quad t \geq 0, \quad (4.18)$$

where f is globally Lipschitz continuous and $g \in \mathcal{C}_b^3$.

- $H = \frac{1}{2}$: Caraballo & Kloeden & Schmalfuss (2011) proves that there exists mean square attractors (L^2 norm), with exponential convergence rate, thus also in the pathwise sense.
- $H > \frac{1}{2}$: LHD & Garrido-Atienza & Neuenkirch & Schmalfuss (2018) proves for evolution equation, criteria quite complicated. Where A -negative definite with $-\lambda$, and F is globally Lipschitz continuous with c_{DF} . $G \in \mathcal{C}^1$, globally Lipschitz continuous with c_{DG} . **But B^H is required to be a small noise!!!**
- $H \in (\frac{1}{3}, \frac{1}{2})$: expect that for small there exists a global pullback attractor \mathcal{A}_g which converges to \mathcal{A} as $\tilde{C}_g \rightarrow 0$. The scheme is similar to finite dimension, but

$$h_t = S(t)h_0 + \int_0^t S(t-u) \left[f(h_u + \mu_u + \mu_u^*) - f(\mu_u + \mu_u^*) \right] du + \int_0^t S(t-u)g(h_u + \mu_u + \mu_u^*)dx_u.$$

And one has to use the norm $\|h\|_{\infty, \beta, \beta, [0_+, t]} = \|h\|_{\infty, [0, t]} + \|(h, h')\|_{\beta, \beta, [0_+, t]}$.

Thank you!