

Fast-slow systems with chaotic noise

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Fast-slow systems

We consider **fast-slow** systems of the form

$$\begin{aligned}\frac{dX}{dt} &= \varepsilon h(X, Y) + \varepsilon^2 f(X, Y) \\ \frac{dY}{dt} &= g(Y),\end{aligned}$$

where $\varepsilon \ll 1$.

$\frac{dY}{dt} = g(Y)$ be some **mildly chaotic** ODE with state space Λ and ergodic invariant measure μ . (**eg.** 3d Lorenz equations.)

$h, f : \mathbb{R}^n \times \Lambda \rightarrow \mathbb{R}^n$ and $\int h(x, y) \mu(dy) = 0$.

Our aim is to find a **reduced equation** $\frac{d\bar{X}}{dt} = F(\bar{X})$ with $\bar{X} \approx X$.

Fast-slow systems

If we rescale to **large time scales** we have

$$\begin{aligned}\frac{dX_\varepsilon}{dt} &= \varepsilon^{-1}h(X_\varepsilon, Y_\varepsilon) + f(X_\varepsilon, Y_\varepsilon) \\ \frac{dY_\varepsilon}{dt} &= \varepsilon^{-2}g(Y_\varepsilon),\end{aligned}$$

We turn X_ε into a random variable by taking $Y(0) \sim \mu$.

The aim is to characterise the **distribution** of the random path X_ε as $\varepsilon \rightarrow 0$.

Why is model reduction important?

- 1** - The reduced model is **lower dimensional** and **less stiff** than the original fast-slow system.
- 2** - Helps the user make **informed guess** when the model is unknown.

Fast-slow systems as SDEs

Consider the simplified **slow** equation

$$\frac{dX_\varepsilon}{dt} = \varepsilon^{-1} h(X_\varepsilon) v(Y_\varepsilon) + f(X_\varepsilon)$$

where $h : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d}$ and $v : \Lambda \rightarrow \mathbb{R}^d$ with $\int v(y) \mu(dy) = 0$.

If we write $W_\varepsilon(t) = \varepsilon^{-1} \int_0^t v(Y_\varepsilon(s)) ds$ then

$$X_\varepsilon(t) = X_\varepsilon(0) + \int_0^t h(X_\varepsilon(s)) dW_\varepsilon(s) + \int_0^t f(X_\varepsilon(s)) ds$$

where the integral is of Riemann-Stieltjes type ($dW_\varepsilon = \frac{dW_\varepsilon}{ds} ds$).

Invariance principle for W_ε

We can write W_ε as

$$W_\varepsilon(t) = \varepsilon \int_0^{t/\varepsilon^2} v(Y(s)) ds = \varepsilon \sum_{j=0}^{\lfloor t/\varepsilon^2 \rfloor - 1} \int_j^{j+1} v(Y(s)) ds$$

The assumptions on Y lead to **decay of correlations** for the sequence $\int_j^{j+1} v(Y(s)) ds$.

For very general classes of chaotic Y , it is known that $W_\varepsilon \Rightarrow W$ in the sup-norm topology, where W is a multiple of Brownian motion.

We will call this class of Y **mildly chaotic**.

What about the SDE?

Since

$$X_\varepsilon(t) = X_\varepsilon(0) + \int_0^t h(X_\varepsilon(s)) dW_\varepsilon(s) + \int_0^t f(X_\varepsilon(s)) ds$$

This suggest a limiting SDE

$$\bar{X}(t) = \bar{X}(0) + \int_0^t h(\bar{X}(s)) \star dW(s) + \int_0^t f(\bar{X}(s)) ds$$

But how should we interpret $\star dW$? Stratonovich? Itô? neither?

For **additive noise** $h(x) = I$
the answer is simple.

Continuity with respect to noise (Sussmann '78)

Consider

$$X(t) = X(0) + \int_0^t dU(s) + \int_0^t f(X(s))ds ,$$

where U is a uniformly continuous path.

The above equation is well defined and moreover $\Phi : U \rightarrow X$ is **continuous** in the sup-norm topology.

Also works in the multiplicative noise case ($h(X)dU$) but only when U is one dimensional.

The simple case (Melbourne, Stuart '11)

If the flow is mildly chaotic ($W_\varepsilon \Rightarrow W$) then $X_\varepsilon \Rightarrow \bar{X}$ in the sup-norm topology, where

$$d\bar{X} = dW + f(\bar{X})ds .$$

In the multiplicative $1d$ noise case, the limit is Stratonovich

$$d\bar{X} = h(\bar{X}) \circ dW + f(\bar{X})ds .$$

The strategy

The solution map takes “irregular path space” to “solution space”

$$\Phi : W_\varepsilon \mapsto X_\varepsilon$$

If this map were **continuous** then we could lift $W_\varepsilon \Rightarrow W$ to $X_\varepsilon \Rightarrow X$.

When the noise is both
multidimensional and
multiplicative, this strategy fails.

Ito, Stratonovich and family

SDEs are very **sensitive** wrt approximations of BM.

Suppose

$$dX = h(X)dW + f(X)dt$$

and define an approximation

$$dX_n = h(X_n)dW_n + f(X_n)dt$$

with some approximation W_n of W .

Taking $n \rightarrow \infty$, X_n might converge to something completely different to X . It all depends on the approximation W_n .

Eg. 1 If W_n is a step function approximation of W , then X_n converges to the **Ito** SDE

$$dX = h(X)dW + f(X)dt$$

Eg. 2 (Wong-Zakai) If W_n is a linear interpolation of W , then X_n converges to the **Stratonovich** SDE

$$dX = h(X) \circ dW + f(X)dt$$

Eg. 3 (McShane, Sussman) If W_n is a higher order interpolation of W , we can get limits which are **neither Ito nor Stratonovich**.

It is not enough to know that

$$W_n \rightarrow BM.$$

We need more information.

Rough path theory (Lyons '97)

Provides a **unified** definition of a DE driven by a noisy path

$$X(t) = X(0) + \int_0^t h(X(s))dU(s) + \int_0^t h(X(s))ds$$

when the dU integral is not well defined.

In addition to U we must be given another path $\mathbb{U} : [0, T] \rightarrow \mathbb{R}^{d \times d}$ which is (formally) an iterated integral

$$\mathbb{U}^{ij}(t) \stackrel{\text{def}}{=} \int_0^t U^i(s)dU^j(s).$$

These extra components tells us how to interpret the **method of integration**.

Rough path theory (Lyons '97)

Given a “rough path” $\mathbf{U} = (U, \mathbb{U})$ we can construct a solution

$$X(t) = X(0) + \int_0^t h(X(s)) d\mathbf{U}(s) + \int_0^t h(X(s)) ds$$

Eg. 1 If $U = W$ and $\mathbb{U} = \int W dW$ is the Ito iterated integral, then the constructed X is the solution to the Ito SDE.

Eg. 2 If $U = W$ and $\mathbb{U} = \int W \circ dW$ is the Stratonovich iterated integral, then the constructed X is the solution to the Stratonovich SDE.

Rough path theory (Lyons '97)

Most importantly (for us) the map

$$\Phi : (U, \mathbb{U}) \mapsto X$$

is an extension of the classical solution map and is **continuous** with respect to the “rough path topology”.

Convergence of fast-slow systems

Returning to the slow variables

$$X_\varepsilon(t) = X_\varepsilon(0) + \int_0^t h(X_\varepsilon(s))dW_\varepsilon(s) + \int_0^t f(X_\varepsilon(s))ds$$

If we let

$$W_\varepsilon^{ij}(t) = \int_0^t W_\varepsilon^i(r)dW_\varepsilon^j(r)$$

then $X_\varepsilon = \Phi(W_\varepsilon, W_\varepsilon)$.

Due to the continuity of Φ , if $(W_\varepsilon, W_\varepsilon) \Rightarrow (W, W)$, then $X_\varepsilon \Rightarrow \bar{X}$, where

$$\bar{X}(t) = \bar{X}(0) + \int_0^t h(\bar{X}(s))dW(s) + \int_0^t f(\bar{X}(s))ds$$

with $W = (W, W)$.

Theorem (K. & Melbourne '14)

If the *fast* dynamics are mildly chaotic, then $(W_\varepsilon, \mathbb{W}_\varepsilon) \Rightarrow (W, \mathbb{W})$ where W is a Brownian motion and

$$\mathbb{W}^{ij}(t) = \int_0^t W^i(s) dW^j(s) + \lambda^{ij} t$$

where the integral is Itô type and

$$\lambda^{ij} = \int_0^\infty \mathbf{E}_\mu \{ v^i(Y(0)) v^j(Y(s)) \} ds .$$

$$\text{Cov}^{ij}(W) = \int_0^\infty \mathbf{E}_\mu \{ v^i(Y(0)) v^j(Y(s)) + v^j(Y(0)) v^i(Y(s)) \} ds$$

Homogenized equations

Corollary

Under the same assumptions as above, the *slow* dynamics $X_\varepsilon \Rightarrow \bar{X}$ where

$$d\bar{X} = h(\bar{X})dW + \left(f(\bar{X}) + \sum_{i,j,k} \lambda^{ij} \partial^k h^i(\bar{X}) h^{kj}(\bar{X}) \right) dt .$$

in Itô form, with $\lambda^{ij} = \int_0^\infty \mathbf{E}_\mu \{ v^i(Y(0)) v^j(Y(s)) \} ds$

$$d\bar{X} = h(\bar{X}) \circ dW + \left(f(\bar{X}) + \sum_{i,j,k} \lambda^{ij} \partial^k h^i(\bar{X}) h^{kj}(\bar{X}) \right) dt$$

in Stratonovich form, with

$\lambda^{ij} = \int_0^\infty \mathbf{E}_\mu \{ v^i(Y(0)) v^j(Y(s)) - v^j(Y(0)) v^i(Y(s)) \} ds .$

General fast-slow systems I

What about the original (much more complicated) fast-slow system?

$$\begin{aligned}\frac{dX_\varepsilon}{dt} &= \varepsilon^{-1} h(X_\varepsilon, Y_\varepsilon) + f(X_\varepsilon, Y_\varepsilon) \\ \frac{dY_\varepsilon}{dt} &= \varepsilon^{-2} g(Y_\varepsilon) .\end{aligned}$$

General fast-slow systems II

The slow variables

$$X_\varepsilon(t) = X_\varepsilon(0) + \int_0^t \varepsilon^{-1} h(X_\varepsilon, Y_\varepsilon) ds + \int_0^t f(X_\varepsilon, Y_\varepsilon) ds$$

can be written in the **product form**

$$X_\varepsilon(t) = X_\varepsilon(0) + \int_0^t H(X_\varepsilon(s)) dW_\varepsilon(s) + \int_0^t H(X_\varepsilon(s)) dV_\varepsilon(s)$$

H is the **evaluation map** (or Dirac distribution) $H(x)\varphi = \varphi(x)$ for $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}^d$ suitably smooth. And $W_\varepsilon, V_\varepsilon$ are the **function valued paths**

$$W_\varepsilon(t) = \varepsilon^{-1} \int_0^t h(\cdot, Y_\varepsilon(s)) ds \quad V_\varepsilon(t) = \int_0^t f(\cdot, Y_\varepsilon(s)) ds$$

General fast-slow systems III

Theorem (K. & Melbourne '14)

If the fast dynamics are mildly chaotic then $X_\varepsilon \Rightarrow \bar{X}$ where

$$d\bar{X} = \sigma(\bar{X})dB + \tilde{a}(\bar{X})dt ,$$

where B is a standard BM on \mathbb{R}^d and

$$\tilde{a}(x) = \int f(x, y) d\mu(y) + \sum_{k=1}^d \mathfrak{B}(h^k(x, \cdot), \partial_k h(x, \cdot))$$

$$\sigma\sigma^T(x) = \mathfrak{B}(h^i(x, \cdot), h^j(x, \cdot)) + \mathfrak{B}(h^j(x, \cdot), h^i(x, \cdot))$$

and \mathfrak{B} is the “integrated autocorrelation” of the fast dynamics

$$\mathfrak{B}(v, w) = \int_0^\infty \mathbf{E}_\mu v(Y(0)) w(Y(s)) ds$$

The real world has feedback

It is more realistic to look fast-slow systems of the form

$$\begin{aligned}\frac{dX_\varepsilon}{dt} &= \varepsilon^{-1}h(X_\varepsilon, Y_\varepsilon) + f(X_\varepsilon, Y_\varepsilon) \\ \frac{dY_\varepsilon}{dt} &= \varepsilon^{-2}g(Y_\varepsilon) + \varepsilon^{\beta-2}g_0(X_\varepsilon, Y_\varepsilon),\end{aligned}$$

for some $\beta \geq 1$. Since the coupling term is of lower order, this is called **weak feedback**.

Back of the envelope: For $\beta > 1$, the reduced model is exactly the same as the the zero feedback case.

For $\beta = 1$, an additional correction term appears, which involves the weak feedback term g_0 .

The real world is infinite dimensional

Many fast-slow models are **PDEs**.

Suppose that $Y_\varepsilon = (Y_\varepsilon^1, Y_\varepsilon^2, \dots)$ is an infinite vector of fast, chaotic variables (possibly coupled). Can we identify a reduced model for $X_\varepsilon = X_\varepsilon(t, x)$ where

$$\partial_t X_\varepsilon = \Delta X_\varepsilon + \varepsilon^{-1} H(X_\varepsilon, Y_\varepsilon) + F(X_\varepsilon, Y_\varepsilon)$$

This is a delicate question, since many natural approximations of noise yield **infinities** in the limiting SPDE.

This is a problem for Hairer's theory of **regularity structures**.

References

- 1 - D. Kelly & I. Melbourne. *Smooth approximations of SDEs*. To appear in **Ann. Probab.** (2014).
- 2 - D. Kelly & I. Melbourne. *Deterministic homogenization of fast slow systems with chaotic noise*. arXiv (2014).
- 3 - D. Kelly. *Rough path recursions and diffusion approximations*. To appear in **Ann. App. Probab.** (2014).

All my slides are on my website (www.dtbkelly.com) **Thank you!**