

# RDS for climate

## Lecture 3: Random time dependent measures

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# Summary and plan of the lecture

- In the first two lectures we have seen several abstract definitions around non autonomous RDS
- we have treated pull-back attractors in some detail
- and we have developed a few examples.
- Today we complete the "Navier-Stokes" example
- and we discussed the difficult topic of time-dependent random measures.
- Finally we start to go back to the main formula relating Weather and Climate.

# The attractor of Navier-Stokes

Recall the finite dimensional example of "Navier-Stokes type":

$$dX_t = (AX_t + B(X_t, X_t)) dt + \sqrt{Q} dW_t$$
$$- \langle Av, v \rangle \geq \nu \|v\|^2$$

$B : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}^d$  bilinear such that

$$\langle B(x, v), v \rangle = 0.$$

Recall we got (for the auxiliary variable  $V_t^{t_0, x_0}(\omega) = X_t^{t_0, x_0}(\omega) - Z_t(\omega)$ )

$$\|V_t^{t_0, x_0}(\omega)\|^2 \leq e^{\int_{t_0}^t (-\nu + C\|Z_r(\omega)\| + \epsilon) dr} \|x_0 - Z_{t_0}(\omega)\|^2$$
$$+ \int_{t_0}^t e^{\int_s^t (-\nu + C\|Z_r(\omega)\| + \epsilon) dr} \frac{C^2}{\epsilon} \|Z_s(\omega)\|^4 ds.$$

Why should  $C\|Z_r(\omega)\|$  be smaller than  $\nu$  most of the time? Not true.

# Extra damping

Instead of using the auxiliary equation

$$dZ_t = AZ_t dt + \sqrt{Q} dW_t$$

let us use the following one:

$$dZ_t^\alpha = (A - \alpha) Z_t^\alpha dt + \sqrt{Q} dW_t.$$

For large  $\alpha > 0$ , the stationary solution is small most of the time (see below). Let us use this new process: for  $V = X - Z^\alpha$  we have

$$\frac{dV_t}{dt} = AV_t + B(V_t + Z_t^\alpha, V_t + Z_t^\alpha) + \alpha Z_t^\alpha$$

$$\frac{1}{2} \frac{d}{dt} \|V_t\|^2 = \langle AV_t, V_t \rangle + \langle B(V_t + Z_t^\alpha, Z_t^\alpha), V_t \rangle + \langle \alpha Z_t^\alpha, V_t \rangle$$

(because  $\langle B(V_t + Z_t^\alpha, V_t), V_t \rangle = 0$ )

$$\frac{1}{2} \frac{d}{dt} \|V_t\|^2 = -\nu \|V_t\|^2 + C \|Z_t^\alpha\|^2 \|V_t\| + C \|Z_t^\alpha\| \|V_t\|^2 + \alpha \|Z_t^\alpha\| \|V_t\|$$

as in the last lecture, with the additional term  $\alpha \|Z_t^\alpha\| \|V_t\|$ .

# Extra damping

Therefore we get

$$\begin{aligned} \|V_t^{t_0, x_0}(\omega)\|^2 &\leq e^{\int_{t_0}^t (-2\nu + 2C\|Z_r^\alpha(\omega)\| + 2\epsilon) dr} \|x_0 - Z_{t_0}^\alpha(\omega)\|^2 \\ &+ \int_{t_0}^t e^{\int_s^t (-2\nu + C\|Z_r^\alpha(\omega)\| + 2\epsilon) dr} \left( \frac{4C^2}{\epsilon} \|Z_s^\alpha(\omega)\|^4 + \frac{4\alpha^2}{\epsilon} \|Z_s^\alpha(\omega)\|^2 \right) ds. \end{aligned}$$

In the sequel we choose  $2\epsilon = \nu$  and simplify this inequality to

$$\begin{aligned} \|V_t^{t_0, x_0}(\omega)\|^2 &\leq e^{\int_{t_0}^t (-\nu + 2C\|Z_r^\alpha(\omega)\|) dr} \|x_0 - Z_{t_0}^\alpha(\omega)\|^2 \\ &+ C_{\nu, \alpha} \int_{t_0}^t e^{\int_s^t (-\nu + C\|Z_r^\alpha(\omega)\|) dr} \left( 1 + \|Z_s^\alpha(\omega)\|^4 \right) ds. \end{aligned}$$

# Extra damping

Now the novelty is that

$$\lim_{t_0 \rightarrow -\infty} \frac{1}{t - t_0} \int_{t_0}^t (-\nu + C \|Z_r^\alpha(\omega)\|) dr = -\nu + C \mathbb{E} [\|Z_0^\alpha(\omega)\|]$$

due to the ergodic theorem, where the limit is understood a.s. in  $\omega$  (we shall be more precise in the dependence on  $t$  below).

It is an easy exercise to prove that

$$\lim_{\alpha \rightarrow \infty} \mathbb{E} [\|Z_0^\alpha(\omega)\|] = 0.$$

Thus choose  $\alpha > 0$  such that

$$-\nu + C \mathbb{E} [\|Z_0^\alpha(\omega)\|] \leq -\frac{2}{3}\nu.$$

It follows that there exists a full measure set  $\tilde{\Omega} \subset \Omega$  such that for all  $\omega \in \tilde{\Omega}$

$$\lim_{t_0 \rightarrow -\infty} \frac{1}{-t_0} \int_{t_0}^0 (-\nu + C \|Z_r^\alpha(\omega)\|) dr =: -\tilde{\nu} \leq -\frac{2}{3}\nu.$$

# Extra damping

Therefore, for all  $\omega \in \tilde{\Omega}$  and all  $t$ ,

$$\lim_{t_0 \rightarrow -\infty} \frac{1}{t - t_0} \int_{t_0}^t (-\nu + C \|Z_r^\alpha(\omega)\|) dr = -\tilde{\nu}.$$

If we had restricted above  $\Omega$  to those  $\omega$  such that  $\|Z_r^\alpha(\omega)\|$  has sub-exponential growth at  $-\infty$ , we deduce

$$\lim_{t_0 \rightarrow -\infty} e^{\int_{t_0}^t (-\nu + 2C \|Z_r^\alpha(\omega)\|) dr} \|x_0 - Z_{t_0}^\alpha(\omega)\|^2 = 0$$

and thus

$$\begin{aligned} & \limsup_{t_0 \rightarrow -\infty} \|V_t^{t_0, x_0}(\omega)\|^2 \\ & \leq R_t(\omega) := C_{\nu, \alpha} \int_{-\infty}^t e^{\int_s^t (-\nu + C \|Z_r^\alpha(\omega)\|) dr} \left(1 + \|Z_s^\alpha(\omega)\|^4\right) ds. \end{aligned}$$

The random variable  $R_t(\omega)$  is finite.

# Extra damping

Going back to

$$U_{\omega}(t_0, t)(x_0) = V_t^{t_0, x_0}(\omega) + Z_t^{\alpha}(\omega)$$

we deduce

$$\limsup_{t_0 \rightarrow -\infty} \|U_{\omega}(t_0, t)(x_0)\|^2 \leq 2R_t(\omega) + 2\|Z_t^{\alpha}(\omega)\|^2.$$

It follows that the ball  $D_{\omega}(t) := B(0, \tilde{R}_t(\omega))$  is absorbing set, with

$$\tilde{R}_t(\omega)^2 = 1 + 2R_t(\omega) + 2\|Z_t^{\alpha}(\omega)\|^2.$$

Then the attractor  $A_{\omega}(t) \subset B(0, \tilde{R}_t(\omega))$  exists.



# Signature of a special perturbative analysis?

It is common to reduce the study of a nonlinear problem like

$$dX_t = (AX_t + \epsilon B(X_t, X_t)) dt + \sqrt{Q} dW_t$$

to the linear case

$$dZ_t = AZ_t dt + \sqrt{Q} dW_t$$

and investigate the difference

$$X_t = Z_t + \dots$$

But when  $\epsilon = 1$ , it is not clear that this work; especially asymptotically in time.

The previous computation shows that we get more informations from

$$X_t = Z_t^\alpha + \dots$$

$$dZ_t^\alpha = (A - \alpha) Z_t^\alpha dt + \sqrt{Q} dW_t.$$

We still do not know whether this has been an accident or it is deep.

# Random time-varying measures

Given a NADS  $(\mathcal{Q}, (\Omega, \mathcal{F}, \mathbb{P}), (X, d), U_\omega(t_0, t))$  (let's drop  $q$  here), we call random time-varying measure a family

$$(\mu_\omega(t))_{t \in \mathbb{R}} \\ \mu_\omega(t)(dx)$$

such that

$$U_\omega(s, t) \mu_\omega(s) = \mu_\omega(t).$$

Assume measurability in  $\omega$ .

# Pull-back convergence to equilibrium

At least in some weak sense, we would like that  $\mu_\omega(t)$  is the limit as  $t_0 \rightarrow -\infty$  of

$$\mu_\omega^{t_0}(t) := U_\omega(t_0, t) \lambda, \quad t \geq t_0$$

for a given probability measure  $\lambda$ , possibly with high geometrical or physical meaning. Let us also formulate it as

$$\lim_{t_0 \rightarrow -\infty} \int_X \phi(U_\omega(t_0, t)(x_0)) \lambda(dx_0) = \int_X \phi(y) \mu_\omega(t)(dy)$$

for all  $\phi \in C_b(X)$ ; with a suitable notion of probabilistic convergence. This is a form of "convergence to equilibrium". In the autonomous deterministic framework it is one of the most difficult and interesting questions. See the outstanding theory of Axiom A hyperbolic attractor, SRB measures and the results about this convergence to equilibrium in that framework.

In the stochastic case there is an additional tool that we now explain.

# Preliminary comments and an example

Before, let us add a few comments.

- There are general results of existence of random time-varying measures  $\mu_\omega(t)$  when the attractor  $A_\omega(t)$  exists, with the property  $\mu_\omega(t)(A_\omega(t)) = 1$  (Crauel 1995).

- One can also prove results of existence of converging subsequences of

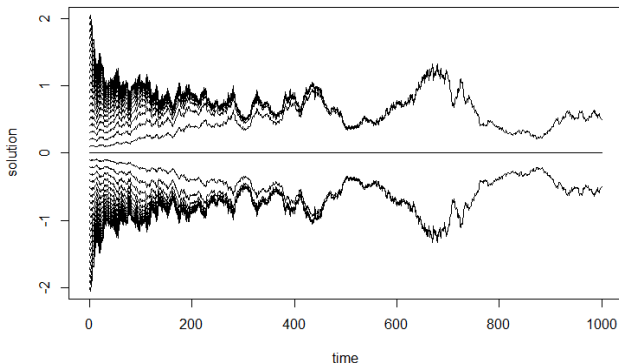
$$\mu_\omega^n(t) := U_\omega(-n, t) \lambda, \quad t \geq t_0.$$

- However, just existence may be poor, it may give us useless measures. Let us see an elementary example.

# Preliminary comments and an example

$$dX_t = (X_t - X_t^3) dt + 0.5 \cdot X_t dW_t$$

$\mu_\omega(t) = \delta_0$  is invariant (but statistically irrelevant)



# Two-point motion

Given a NADS (let us drop  $q$  here for simplicity of notations)

$$(\mathcal{Q}, (\Omega, \mathcal{F}, \mathbb{P}), (X, d), U_\omega(t_0, t))$$

we call two-point motion the map

$$U_\omega^{(2)}(t_0, t) : X \times X \rightarrow X \times X$$

simply defined as

$$U_\omega^{(2)}(t_0, t)(x_0, x'_0) = (U_\omega(t_0, t)(x_0), U_\omega(t_0, t)(x'_0)).$$

In a sense, it is just a reduced version of the map  $U_\omega(t_0, t)$  itself: we observe the motion of two different initial conditions under the same values of  $\omega$ , instead of observing the motion of the full page space.

# Ergodicity of the two-point motion

Let us take advantage of the random aspect of the parameter  $\omega$  and define the map

$$\mathcal{P}_2(t_0, t) : C_b(X \times X) \rightarrow C_b(X \times X)$$
$$(\mathcal{P}_2(t_0, t) F)(x_0, x'_0) := \mathbb{E} \left[ F \left( U^{(2)}(t_0, t)(x_0, x'_0) \right) \right].$$

## Definition

We say that the two-point motion is pull-back exponentially ergodic if there exists a family of probability measures  $\left( \rho^{(2)}(t) \right)_{t \in \mathbb{R}}$  on Borel sets of  $X \times X$  such that the following holds: for every  $t \in \mathbb{R}$  and  $F \in C_b(X \times X)$  there exists  $C_{t,F}, \nu_{t,F} > 0$  such that

$$\left| \langle \mathcal{P}_2(t_0, t) F, \rho_0 \rangle - \langle F, \rho^{(2)}(t) \rangle \right| \leq C_{t,F} e^{-\nu_{t,F}(t-t_0)}$$

for all  $t_0 \leq t$  and  $\rho_0 \in \text{Pr}(X \times X)$ .

# White noise random dynamical systems

Assume that on  $(\Omega, \mathcal{F}, \mathbb{P})$  there is a filtration  $(\mathcal{F}_t)_{t \in \mathbb{R}}$  such that  $U_\omega(t_0, t)$  is  $\mathcal{F}_t$ -measurable and  $\mathcal{F}_{t_0}$ -independent, for every  $t_0 < t$ . In this case we say that

$$(\mathcal{Q}, (\Omega, \mathcal{F}, \mathbb{P}), (X, d), U_\omega(t_0, t))$$

is a *white noise* random dynamical system (or NADS).

This property is true for solution of SDEs (and SPDEs) driven by an  $(\mathcal{F}_t)_{t \in \mathbb{R}}$ -Brownian motion (or more generally an  $(\mathcal{F}_t)_{t \in \mathbb{R}}$ -Lévy processes) since the solution map on  $[t_0, t]$  depends only on the increments of the Brownian motion on  $[t_0, t]$ , which are  $\mathcal{F}_t$ -measurable but also  $\mathcal{F}_{t_0}$ -independent.



# Towards the Physical random measure

## Lemma

*Consider a white noise NADS. Assume that the two-point motion is pull-back exponentially ergodic. Then, given  $t \in \mathbb{R}$ ,  $\lambda \in \text{Pr}(X)$  and  $\phi \in C_b(X)$ , the family of random variables*

$$\left( \int_X \phi(U_\omega(t_0, t)(x_0)) \lambda(dx_0) \right)_{t_0 \in \mathbb{R}}$$

*is Cauchy in Probability as  $t_0 \rightarrow -\infty$ .*

## Theorem

*Under the previous assumptions there is a (unique) measure  $\mu_\omega(t)$  such that, for  $t \in \mathbb{R}$ ,  $\lambda \in \text{Pr}(X)$  and  $\phi \in C_b(X)$ , in Probability,*

$$\lim_{t_0 \rightarrow -\infty} \int_X \phi(U_\omega(t_0, t)(x_0)) \lambda(dx_0) = \int_X \phi(y) \mu_\omega(t)(dy).$$

# Proof of the Lemma

Given  $t'_0 < t_0$ ,

$$\begin{aligned} & \mathbb{E} \left[ \left( \int_X \phi(U(t_0, t)(x_0)) \lambda(dx_0) - \int_X \phi(U(t'_0, t)(x_0)) \lambda(dx_0) \right)^2 \right] \\ &= -2 \int_{X^2} \mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t'_0, t)(x'_0))] \lambda(dx_0) \lambda(dx'_0) \\ & \quad + \langle \mathcal{P}_2(t_0, t)(\phi \otimes \phi), \lambda \otimes \lambda \rangle + \langle \mathcal{P}_2(t'_0, t)(\phi \otimes \phi), \lambda \otimes \lambda \rangle. \end{aligned}$$

By assumption we know that  $\langle \mathcal{P}_2(t_0, t)(\phi \otimes \phi), \lambda \otimes \lambda \rangle$  converges to  $\langle \phi \otimes \phi, \rho^{(2)}(t) \rangle$  exponentially fast, as  $t_0 \rightarrow -\infty$ . The same is true for  $\langle \mathcal{P}_2(t'_0, t)(\phi \otimes \phi), \lambda \otimes \lambda \rangle$  for  $t'_0 \rightarrow -\infty$ . It remains to prove that

$$\int_{X^2} \mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t'_0, t)(x'_0))] \lambda(dx_0) \lambda(dx'_0)$$

is close to  $\langle \phi \otimes \phi, \rho^{(2)}(t) \rangle$  when both  $t'_0 < t_0$  are close enough to  $-\infty$ .

We have (using the white noise property)

$$\begin{aligned} & \mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t'_0, t)(x'_0))] \\ = & \mathbb{E} [\mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t_0, t)U(t'_0, t_0)(x'_0)) | \mathcal{F}_{t_0}]] \\ = & \mathbb{E} [\mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t_0, t)(y_0))]_{y_0=U(t'_0, t_0)(x'_0)}] \\ = & \mathbb{E} [(\mathcal{P}_2(t_0, t)(\phi \otimes \phi))(x_0, y_0)_{y_0=U(t'_0, t_0)(x'_0)}]. \end{aligned}$$

Hence

$$\begin{aligned} & \int_{X^2} \mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t'_0, t)(x'_0))] \lambda(dx_0) \lambda(dx'_0) \\ = & \mathbb{E} \left[ \int_{X^2} (\mathcal{P}_2(t_0, t)(\phi \otimes \phi))(x_0, y_0)_{y_0=U(t'_0, t_0)(x'_0)} \lambda(dx_0) \lambda(dx'_0) \right]. \end{aligned}$$

# Proof (cont.)

Now, writing  $\phi_2$  for  $\phi \otimes \phi$ ,

$$\begin{aligned} & \int_{X^2} (\mathcal{P}_2(t_0, t) \phi_2)(x_0, y_0)_{y_0 = U_\omega(t'_0, t_0)(x'_0)} \lambda(dx_0) \lambda(dx'_0) \\ &= \int_{X^2} (\mathcal{P}_2(t_0, t) \phi_2)(x_0, y_0) \lambda(dx_0) \nu_\omega(t'_0, t_0)(dy_0) \end{aligned}$$

where  $\nu_\omega(t'_0, t_0) = U_\omega(t'_0, t_0) \lambda$  and

$$\left| \langle \mathcal{P}_2(t_0, t) \phi_2, \lambda \otimes \nu_\omega(t'_0, t_0) \rangle - \langle \phi_2, \rho^{(2)}(t) \rangle \right| \leq C_{t, \phi_2} e^{-\nu_{t, \phi_2}(t-t_0)}$$

hence

$$\begin{aligned} & \left| \int_{X^2} \mathbb{E} [\phi(U(t_0, t)(x_0)) \phi(U(t'_0, t)(x'_0))] \lambda(dx_0) \lambda(dx'_0) - \langle \phi_2, \rho^{(2)}(t) \rangle \right| \\ & \leq C_{t, \phi_2} e^{-\nu_{t, \phi_2}(t-t_0)}. \end{aligned}$$

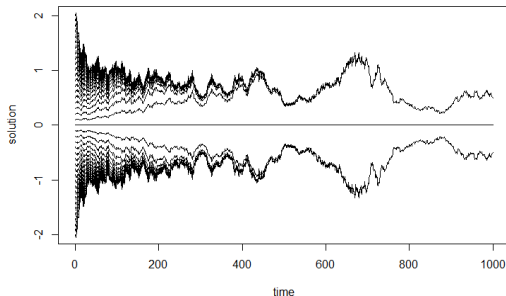
# Back to the example

Presumably one can prove that

$$dX_t = (X_t - X_t^3) dt + 0.5 \cdot X_t dW_t$$

on the metric space  $\mathbb{R} \setminus \{0\}$  is pull-back exponentially ergodic with unique

$$\mu_\omega(t) = \frac{1}{2} \delta_{x_t^-(\omega)} + \frac{1}{2} \delta_{x_t^+(\omega)}$$



# Summary

- We have seen that in special cases there is hope to have a unique measure

$$\left( \mu_{q,\omega}(t) \right)_{t \in \mathbb{R}}$$

such that

$$U_{q,\omega}(s, t) \mu_{q,\omega}(s) = \mu_{q,\omega}(t)$$

given by the pull-back of significant measures.

- Such family  $\left( \mu_{q,\omega}(t) \right)_{t \in \mathbb{R}}$  represents what we observe statistically in the long run, independently of the initial conditions.
- Using this family we would like to understand the problem

$$\lim_{\epsilon \rightarrow 0} \int_{-\infty}^{\infty} \int_X \psi(t, y) \mu_{q,\omega}^{\epsilon}(s)(dy) ds = \int_{-\infty}^{\infty} \int_X \psi(t, x) \mu_q(t)(dx).$$

posed in the first lecture.

- Meanwhile, in the Problem Sessions, we are developing the necessary technical elements to treat such a limit.
- The problem is that  $\mu_{q,\omega}^\epsilon(t)$  is extremely implicit. We do not have so much concrete information on it to investigate the limit.
- One property is

$$\mu_{q,\omega}^\epsilon(t)(A_{q,\omega}^\epsilon(t)) = 1.$$

Can we use the random attractor to prove tightness properties of  $\mu_{q,\omega}^\epsilon(t)$ ?

- This is the question we pose for the third Problem Session and we shall use the result in the next Lecture, n.4.

# Self-criticism and open questions

- The condition of pull-back exponentially ergodicity of the two-point motion is poorly investigated: more research on it would be interesting.
- It is perhaps very restrictive and special: it implies that  $\mu_\omega(t)$  is independent of  $\lambda$ ! (Similar proof.)
- For Axiom A hyperbolic attractors, there is a unique  $\mu$  with nice properties but there are infinitely many others which are unstable and less relevant (delta Dirac at unstable fixed points and so on).
- Thus, perhaps, the previous theory "homogenizes" too much. Or maybe not, maybe that is "the rule" for truly stochastic models.



# Thank you for your attention

