

1. Random dynamical systems: important definitions and examples

1.1 Definition of a random dynamical system

Def.: $(\mathcal{N}, \mathcal{F}, \mathbb{P})$ probability space,
 (X, \mathcal{B}) measurable space, $\bar{\Pi} \in \{\mathbb{Z}, \mathbb{Z}_+^+, \mathbb{N}, \mathbb{N}_+^+\}$

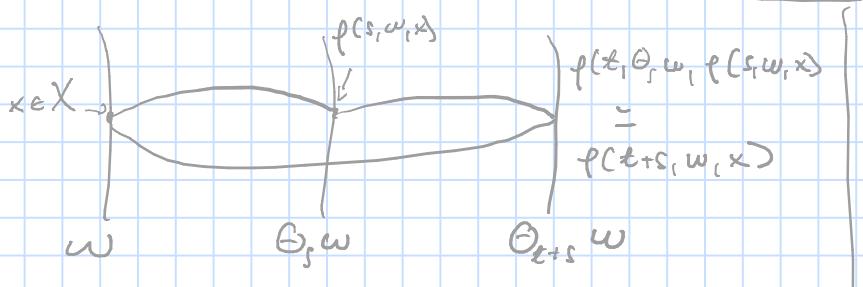
① A random dynamical system (RDS) is a pair (Θ, φ) with:

- measurable $\Theta: \bar{\Pi} \times \mathcal{N} \rightarrow \mathcal{N}$, $(t, \omega) \mapsto \Theta_t \omega$ s.t.

(measure-preserving)
 PS (i) $\Theta_0 = \text{id}$, $\Theta_{t+s} = \Theta_t \circ \Theta_s$, $\forall s, t \in \bar{\Pi}$
 (ii) $\mathbb{P}(A) = \mathbb{P}(\Theta_t^{-1} A)$, $\forall A \in \mathcal{F}, t \in \bar{\Pi}$

- meas. $\varphi: \bar{\Pi} \times \mathcal{N} \times X \rightarrow X$, $(t, \omega, x) \mapsto \varphi(t, \omega, x)$

(cocycle) (i) $\varphi(0, \omega, \cdot) = \text{id}$ $\forall \omega \in \mathcal{N}$
 (ii) $\varphi(t+s, \omega, \cdot) = \varphi(t, \Theta_s \omega, \varphi(s, \omega, \cdot))$
 $\forall \omega \in \mathcal{N}, t, s \in \bar{\Pi}$



(2) If X topological space with $\mathcal{B} = \mathcal{B}(X)$ (Worcl G-alg.)

and

$$f(\cdot, w, \cdot) : \mathbb{T} \times X \rightarrow X, (t, x) \mapsto f(t, w, x)$$

is continuous for all $w \in \mathbb{N}$:

Then the RNS (G, f) is called **continuous**.

(3) If X is a smooth (C^∞) manifold (e.g. \mathbb{R}^d)

$$f(t, w, \cdot) : X \rightarrow X, x \mapsto f(t, w, x), \text{ is } C^k.$$

Then the RNS (G, f) is called C^k .

1.2. Generating RNS in discrete time

Propn.: (Θ, f) RNS on X , $\mathbb{T} \in \{\mathbb{Z}_0^+, \mathbb{Z}\}$

(1) $\mathbb{T} = \mathbb{Z}_0^+$: Consider

$$\boxed{\psi(w) := f(1, w) : X \rightarrow X.}$$

We obtain

$$f(n, w) = \begin{cases} \psi(\Theta_{n-1} w) \circ \dots \circ \psi(w), & n \geq 1. \\ \text{id}, & n=0. \end{cases}$$

(2) $\mathbb{T} = \mathbb{Z}$: Additionally, we have

$$\boxed{f(-1, w) = f(1, \Theta_{-1} w)^{-1} = \psi(\Theta_{-1} w)^{-1}}$$

such that $\psi(w)$ invertible and

$$f(n, w) = \psi(\Theta_n w)^{-1} \circ \dots \circ \psi(\Theta_{-1} w)^{-1}, n \leq -1.$$

Proof: Application of cocycle property. \square

Continuity/Differentiability of the NBS follow from
— II — of

$x \mapsto \psi(\omega)x$ $\mathbb{H}^{\text{w.t.}}$ (and $x \mapsto \psi(\omega)^{-1}x$ if $\Pi = \mathbb{Z}$).

1.2.1 Examples

(1) Linear NBS as product of random matrices: $X = h^d$, NBS linear (in x):

$$n \geq 1: \quad \varphi(n, \omega) = A_{n-1}(\omega) \cdots A_0(\omega), \quad A_\varepsilon(\omega) = A(\theta_\varepsilon \omega)$$

for some meas. $A: \mathbb{N} \rightarrow \mathbb{R}^{d \times d}$.

$\rightarrow \Pi = \mathbb{Z}$:

$$n \leq -1: \quad \varphi(n, \omega) = A_n(\omega)^{-1} \cdots A_{-1}(\omega)^{-1}, \quad A_\varepsilon(\omega) = A(\theta_\varepsilon \omega).$$

(1*) Herman's example:

$$A(\omega) = \begin{bmatrix} G & 0 \\ 0 & G^{-1} \end{bmatrix} \begin{bmatrix} \cos 2\pi\omega & -\sin 2\pi\omega \\ \sin 2\pi\omega & \cos 2\pi\omega \end{bmatrix},$$

• $\omega \in S^1 (\simeq [0, 1])$

$(\theta_n := \theta^n)$ • $\Theta: S^1 \rightarrow S^1$ irrational rotation,

keeping Lebesgue meas. invariant

(\rightarrow later: pos. first Lyapunov exponent)

(2) Chaos game for Cantor set:

$$x \in X = [0, 1]: \quad T_0(x) = \frac{x}{3}, \quad T_1(x) = \frac{2+x}{3}$$

Random switching as an RDS:

- $\Delta = \{0, 1\}$, $\mathcal{R} = \Delta^{\mathbb{N}} := \{ \omega = (\omega_n)_{n=0}^{\infty} \mid \omega_n \in \Delta \}$
- **Cylinder sets** $C_{i_0, \dots, i_m} = \{ \omega \in \mathcal{R} \mid \omega_n = i, n=0, \dots, m \}$
- Probability measure ν : $\nu(0) = \mu_0$, $\nu(1) = \mu_1 = 1 - \mu_0$
- $(\mathcal{R}, \mathcal{B}(\mathcal{R}))$ endowed with $\text{IP} = \nu^{\mathbb{N}}$
given by $\text{IP}(C_{i_0, \dots, i_m}) = \mu_{i_0} \cdots \mu_{i_m}$
(\rightarrow Carathéodory extension)
- Shift map $\Theta : \mathcal{R} \rightarrow \mathcal{R}$
 $\Theta(\omega_n)_{n=0}^{\infty} = (\omega_{n+1})_{n=0}^{\infty}$
 \rightarrow Cocycle $\varphi : \mathbb{Z}_0^+ \times X \times \mathcal{R} \rightarrow X$ as
 $\varphi(0, \omega, x) = x$, $\varphi(n, \omega, x) = T_{i_{n-1}} \circ \dots \circ T_{i_0}(x)$,
where $\omega = (\omega_n)_{n=0}^{\infty}$.

Construction can be generalized for random maps drawn from compact metric space Δ and $\mathcal{R} = \Delta^{\mathbb{Z}}$, i.e. two-sided time.

1.2.2

Relation to Markov chains

Theorem: RDS (Θ, γ) generated by $\psi(\omega)$

The orbit (x_n) given by

$$x_{n+1} = \psi(\Theta_n \omega) x_n, \quad x_0 = x \in X,$$

is a **Markov chain** with

$$\text{P}(x, B) = \text{IP}(\{\omega : \psi(\omega)x \in B\}) \quad \forall B \in \mathcal{B}.$$

Proof: See notes. □

Remark:

Reverse: Markov chain \rightarrow RDS

true under mild conditions (see Kifer '86)
(in general, uniqueness not guaranteed)

1.3 RDS from SDEs

Consider the Stratonovich SDE (midpoint rule)

$$(SDE) \quad dX_t = f(X_t)dt + \sum_{j=1}^m g_j(X_t) \circ dW_t^j, \quad X_0 = x \in \mathbb{H}^d$$

where $f \in C_b^{2,\delta}$, $g_j \in C_b^{2+1,\delta}$, $\delta \in \mathbb{N}_1$, $\delta \in [0,1]$,
and W_t^j are independent Brownian motions.

1.3.1 Brownian motion as dynamical system

Def. and Prop.: Wiener space $(\mathcal{N}, \mathcal{F}, \mathbb{P})$

is given as follows:

$$\textcircled{1} \quad \mathcal{N} = C_0(\mathbb{R}, \mathbb{H}^d) = \{ \omega: \mathbb{R} \rightarrow \mathbb{H}^d \text{ cont.}: \omega(0) = 0 \}$$

\mathcal{N} is endowed with the compact-open

topology given by the complete metric

$$s(\omega, \tilde{\omega}) := \sum_{n=1}^{\infty} \frac{1}{2^n} \frac{\|\omega - \tilde{\omega}\|_n}{1 + \|\omega - \tilde{\omega}\|_n},$$

$$\|\omega - \tilde{\omega}\| := \sup_{|t| \leq n} \|\omega(t) - \tilde{\omega}(t)\|.$$

②. $\mathcal{F} = \mathcal{B}(\mathbb{R})$, the Borel σ -algebra on $(\mathbb{R}, \mathcal{F})$.

③. Wiener measure \mathbb{P} on $(\mathbb{R}, \mathcal{F})$:

$$\mathbb{P}(\{\omega \in \mathbb{R} : w_1(t) \leq x_1, \dots, w_d(t) \leq x_d\})$$

$$= \frac{1}{(2\pi t)^{d/2}} \int_{-\infty}^{x_1} \cdots \int_{-\infty}^{x_d} e^{-\|\gamma\|^2/2t} d\gamma_1 \cdots d\gamma_d,$$

for all $x \in \mathbb{R}^d$.

($(W_t^1(\omega), \dots, W_t^d(\omega))^T := w(t)$ are independent BMs)

④. $\mathcal{F}_s^t \subset \mathcal{F}$ (sub- σ -algebra \rightarrow filtration)

generated by $\{w(u) - w(v) \mid s \leq v \leq u \leq t\}$.

⑤. Family of shift maps $(\Theta_t)_{t \in \mathbb{R}}$ on $(\mathbb{R}, \mathcal{F}, \mathbb{P})$

$$\Theta_t w(\cdot) = w(t + \cdot) - w(t)$$

is measure-preserving and ergodic

($t \in \mathbb{R}, \Theta_t^{-1}(A) \subseteq A \Rightarrow \mathbb{P}(A) = 0$ or $\mathbb{P}(A) = 1$)

Proof: See notes and Schurz '10. \square

1.3.2

SDE solutions as cocycles

Theorem: Consider (SDEs) as above:

there is a unique meas. map $(t, \omega, x) \mapsto \varphi(t, \omega, x)$
such that

- (θ, φ) is a C^k RDS
- $\varphi(t, \cdot, x)$ solves (SDE).

Proof: See Arnold/Schutzbaw '85.

Sketch:

① Establish that solutions of (SDE) give
 C^k flow of diffeomorphisms ($\varphi_{st}(\omega) := \varphi_s(\omega, \cdot)$)

(ZPF) $\varphi_{rt}(\omega) \circ \varphi_{rs}(\omega) = \varphi_{st}(\omega)$, $\varphi_{ss}(\omega) = \text{id}$.

(Check: $dX_t = dW_t$ generates C^∞ flow

$x \rightsquigarrow \varphi_{st}(\omega, x) = x + W_t(\omega) - W_s(\omega)$ [cf. Kunita '90].

② $\varphi_{rt}(\omega) = \varphi_{0, t-r}(\theta_r \omega)$ for almost all (ω)
(Check for $dX_t = dW_t$)

$\varphi(t, \cdot) := \varphi_{0t}(\cdot)$, replace t, r, s in (ZPF) by $t+s, s, 0$

$\rightarrow \varphi(t+s, \omega) = \varphi(t, \theta_s \omega) \circ \varphi(s, \omega)$ for fixed $s \in \mathbb{N}$,
all $t \in \mathbb{N}$ IP-a.s.

\rightarrow Crude cocycle: Exceptional null set may depend on s .

③ Perfection of cocycle via abstract argument, using Haar measures. \square

1.4

Invariant measures and corresp-

ondence theorem

RDS (Θ, φ) with time $\bar{\Pi}$.

→ Skew-product flow

$$\overline{\Theta}_t (\omega, x) = (\Theta_t \omega, \varphi(t, \omega, x))$$

For measure μ , $T^* \mu(\cdot) = \mu(T^{-1}(\cdot))$ push forward

Def.: Prob. meas. μ on $(\mathcal{N} \times X, \mathcal{F} \otimes \mathcal{B})$ is invariant for (Θ, φ) if

(i) $\overline{\Theta}_t^* \mu = \mu$ for all $t \in \bar{\Pi}$.

(ii) $\mu(d\omega, dx) = \mu_\omega(dx) \mathbb{P}(d\omega)$,

where $\omega \mapsto \mu_\omega$ is a random/sample measure.

- μ_ω is a prob. meas. on X a.s.
- $\omega \mapsto \mu_\omega(B)$ meas. for all $B \in \mathcal{B}$.

Remark: For X Polish space (complete, separable, metric space):

μ invariant for (Θ, φ) iff

$$\varphi(t, \omega, \cdot)^* \mu_\omega = \mu_{\Theta_t \omega}, \text{ a.s. } \forall t \in \bar{\Pi}.$$

Now, consider RDS (Θ, φ) with $\mathcal{F}_v^\vee \subset \mathcal{F}$ generated by $\varphi(s, \Theta_s \omega, x)$, for $x \in X$, $t, s \in \bar{\Pi}$ with $0 \leq s \leq v$ and $0 \leq t \leq v-s$.

$\mathcal{F}_{-v}^\vee, \mathcal{F}_v^\vee$ natural extensions.

Def. RNS (G, \mathcal{F}) is called Markov if $\mathcal{F}_{-\infty}^0$ (the past) and \mathcal{F}_0^∞ (the future) are independent.

Note: Markov RNS \Rightarrow Markov process;
 Markov process and RNS (e.g. (SDE))
 \Rightarrow Markov RNS

Def.: Invariant meas. μ for (G, \mathcal{F}) is Markov if $\{\mu_w\}$ is $\mathcal{F}_{-\infty}^0$ -meas.,

Theorem [Correspondence of measures]

(G, φ) Markov RNS on Polish space X :

(a) If μ invar. Markov meas. with samples μ_w , then $\varrho = \mathbb{E}[\mu_w]$ stationary for the Markov process ($P_t^* \varrho = \varrho$ for dual semi-group)
 \oplus : μ' s.t. $\mathbb{E}[\mu_w] = \mathbb{E}[\mu'_w] \Rightarrow \mu = \mu'$

(b) ϱ stationary: Then for $t_n \rightarrow \infty$, almost surely
 $\mu_w = \lim_{n \rightarrow \infty} f(t_n, \Theta_{-t_n} w, \cdot)^*$ (e) weakly.

$\rightarrow \mu$ invariant Markov meas. and
 $\varrho = \mathbb{E}[\mu_w]$.

Summarizing: one-to-one correspondence between stationary and invariant Markov meas..

Proof: See Lecture Notes (or directly
[Kuksin / Shirikyan '12]). \square

Note: $E[\mu_\omega(\cdot) | \mathcal{F}_0^\infty] = E[\mu_\omega(\cdot)] = g(\cdot)$
 $\Rightarrow E[\mu(\cdot) | \mathcal{F}_0^\infty] = (\mathbb{P} \times g)(\cdot).$

Example: O-U process:

(O-U) $dX_t = -\gamma X_t dt + \sigma dW_t$, $\gamma > 0$.
 Stationary $p(dx) = \underbrace{\int e^{-\frac{\gamma x^2}{2\sigma^2}} dx}_{\text{normalization}}$

\rightarrow gives invariant Markov meas. μ
 for Markov RNS induced by (OU).