

SELF-INTERACTING DIFFUSIONS II: CONVERGENCE IN LAW

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Received 28 May 2002, accepted 26 February 2003

ABSTRACT. – This paper concerns convergence in law properties of self-interacting diffusions on a compact Riemannian manifold.

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RÉSUMÉ. – Cet article étudie les propriétés de convergence en loi des diffusions interagissantes sur une variété riemannienne compacte.

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1. Introduction

Self interacting diffusions (as considered here) are continuous time stochastic processes living on a Riemannian manifold M which can be typically described as solutions to a stochastic differential equation (SDE) of the form

$$dX_t = \sum_{\alpha} F_{\alpha}(X_t) \circ dB_t^{\alpha} - \frac{1}{t} \left(\int_0^t \nabla V_{X_s}(X_t) ds \right) dt, \quad (1)$$

where $(B^{\alpha})_{\alpha}$ is a family of independent Brownian motions, $(F_{\alpha})_{\alpha}$ is a family of smooth vector fields on M such that $\sum_{\alpha} F_{\alpha}(F_{\alpha}f) = \Delta f$ (for $f \in C^{\infty}(M)$), where Δ denotes the Laplacian on M and $V_u(x)$ a “potential” function.

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These processes are characterized by the fact that the drift term in Eq. (1) depends both on the position of the process X_t , and its empirical occupation measure up to time t :

$$\mu_t = \frac{1}{t} \int_0^t \delta_{X_s} ds. \quad (2)$$

The asymptotic behavior of $\{\mu_t\}$ is the subject of a recent paper by Benaïm, Ledoux and Raimond [1]. This paper provides tools and results which allow to describe the long term behavior of $\{\mu_t\}$ in terms of the long term behavior of a certain deterministic semi-flow $\{\Psi_t\}_{t \geq 0}$ defined on the space of probability measure on M . For instance, there are situations (depending on the shape of V) in which $\{\mu_t\}$ converges almost surely to an equilibrium point μ^* of Ψ and other situations where the limit set of $\{\mu_t\}$ coincides almost surely with a periodic orbit for Ψ (see the examples in Section 4 of [1] and below in Section 7). In the simple case where μ_t converges to μ^* one expects $(X_{t+s}, s \geq 0)$ to behave like a homogeneous diffusion of generator

$$L_{\mu^*} = \frac{1}{2} \Delta + \langle \nabla V_{\mu^*}, \nabla \rangle,$$

where $V_{\mu^*}(x) = \int V_y(x) \mu^*(dy)$ and $\langle \cdot, \cdot \rangle$ denotes the Riemannian inner product on M . The purpose of this note is to address this type of question.

In Section 2, following [1], self-interacting diffusions on a smooth compact manifold are defined. In Section 3, the basic tool of this paper is presented, namely the Girsanov transform.

In Section 4, we show that on the event “ μ_t converges towards μ^* ”, the law of $(X_{t+u}, u \geq 0)$ given $\mathcal{B}_t = \sigma(X_s, s \leq t)$ is asymptotically equal to the law of the diffusion with generator L_{μ^*} and initial condition X_t .

In Section 5, we show that the law of $X_{t+s(t)}$ given \mathcal{B}_t is asymptotically equal to $\Pi(\mu_t)$, the invariant probability measure of the diffusion with generator L_{μ_t} ; provided $s(t) \rightarrow \infty$ at a convenient rate. Moreover the law of X_t given $\tilde{\Omega} = \{\mu_t \rightarrow \mu^*\}$ converges towards $E[\mu^* | \tilde{\Omega}]$. In particular, when $P(\tilde{\Omega}) = 1$, X_t converges in law towards $E[\mu^*]$.

Section 6 generalizes results of Section 5 to the law of the process $(X_{t+s(t)+v}, v \geq 0)$.

In Section 7, examples developed in [1] and [2], for which μ_t converges a.s. are presented.

2. Background and notation

The notation and definitions here are from [1].

Throughout we let M denote a d -dimensional, compact connected smooth (C^∞) Riemannian manifold. Without loss of generality (see Nash [4]) we shall assume that M is isometrically embedded in \mathbb{R}^N . We denote $C^r(M)$, $0 \leq r \leq \infty$, the space of C^r real valued functions on M .

Given a metric space E we let $\mathcal{P}(E)$ denote the space of Borel probability measures on E equipped with the topology induced by the weak convergence. Recall that a sequence

$\{\mathbb{P}_n\}_{n \geq 0}$ of Borel probability measures on E converges weakly to \mathbb{P} provided

$$\lim_{n \rightarrow \infty} \int f d\mathbb{P}_n = \int f d\mathbb{P} \tag{3}$$

for every bounded and continuous function $f : E \rightarrow \mathbb{R}$. When E is compact (e.g., $E = M$), $\mathcal{P}(E)$ is a compact metric space.

Throughout we assume given a measurable mapping

$$\begin{aligned} V : M \times M &\rightarrow \mathbb{R}, \\ (u, x) &\mapsto V(u, x) = V_u(x). \end{aligned} \tag{4}$$

We furthermore assume that for all $u \in M$, $V_u : M \rightarrow \mathbb{R}$ is a C^1 function whose first derivatives are bounded (in the variables u and x). For $\mu \in \mathcal{P}(M)$ we let $V_\mu \in C^1(M)$ denote the function defined by

$$V_\mu(x) = \int_M V(u, x) \mu(du), \tag{5}$$

and L_μ the operator defined on $C^\infty(M)$ by

$$L_\mu f = \frac{1}{2} \Delta f - \langle \nabla V_\mu, \nabla f \rangle, \tag{6}$$

where $\langle \cdot, \cdot \rangle$, ∇ and Δ stand, respectively, for the Riemannian inner product, the associated gradient and Laplacian on M .

We let Ω denote the space of continuous paths $w : \mathbb{R}_+ \rightarrow M$, equipped with the topology of uniform convergence on compact intervals; $\mathcal{B} = \mathcal{B}(\Omega)$ the Borel σ -field of Ω , X_t the M -valued random variable defined by $X_t(w) = w(t)$; and \mathcal{B}_t the σ -field generated by the random variables $\{X_s : 0 \leq s \leq t\}$.

Since Ω is polish, $\mathcal{P}(\Omega)$ equipped with the weak convergence is metrizable. A distance d on $\mathcal{P}(\Omega)$ is given by

$$d(\mathbb{P}, \mathbb{Q}) = \sum_{n=1}^{\infty} 2^{-n} \left| \int Z_n d\mathbb{P} - \int Z_n d\mathbb{Q} \right| \tag{7}$$

for \mathbb{P} and \mathbb{Q} in $\mathcal{P}(\Omega)$ where $Z_n : \Omega \rightarrow \mathbb{R}$ is continuous, \mathcal{B}_n -measurable, and $\{Z_n; n \geq 1\}$ is dense in $\{Z \in C^0(\Omega); \|Z\|_\infty \leq 1\}$.

For $r > 0$, $\mu \in \mathcal{P}(M)$ and $w \in \Omega$, the empirical occupation measure of w with initial weight r and initial measure μ is the sequence $\{\mu_t(r, \mu, w) \in \mathcal{P}(M) : t \geq 0\}$ defined by

$$\mu_t(r, \mu, w) = \frac{1}{r+t} \left(r\mu + \int_0^t \delta_{w(s)} ds \right), \tag{8}$$

where $\int_0^t \delta_{w(s)} ds(A) = \int_0^t \mathbf{1}_A(w(s)) ds$, for every Borel set $A \subset M$. In the following we will denote by $\mu_t(r, \mu)$ the $\mathcal{P}(M)$ -valued random variable $w \mapsto \mu_t(r, \mu, w)$.

A self-interacting diffusion associated to V is a family

$$\{\mathbb{P}_{x,r,\mu}: x \in M, r > 0, \mu \in \mathcal{P}(M)\} \subset \mathcal{P}(\Omega) \tag{9}$$

such that

- (i) $\mathbb{P}_{x,r,\mu}(X_0 = x) = 1$.
- (ii) For all $f \in C^\infty(M)$,

$$M_t^f = f(X_t) - f(x) - \int_0^t (L_{\mu_s(r,\mu)} f)(X_s) ds$$

is a $\mathbb{P}_{x,r,\mu}$ -martingale relative to $\{\mathcal{B}_t: t \geq 0\}$.

Existence and uniqueness of the self-interacting diffusion associated to V is proved in [1], Proposition 2.5. More precisely, it is shown in this paper that $\mathbb{P}_{x,r,\mu}$ can be obtained as the law of $\{X_t\}$, a solution (unique in law) of the following SDE on M :

$$dX_t = \sum_{i=1}^N F_i(X_t) \circ dB_t^i - \nabla V_{\mu_t(r,\mu)}(X_t) dt, \quad X_0 = x, \tag{10}$$

where $(F_1(x), \dots, F_N(x))$ denote the orthogonal projection of the canonical basis (e_1, \dots, e_N) of \mathbb{R}^N on $T_x M$ and $B_t = (B_t^1, \dots, B_t^N)$ is an N -dimensional Brownian motion.

For $x \in M$ and $\mu \in \mathcal{P}(M)$ we let $\mathbb{P}_{x,\mu} \in \mathcal{P}(\Omega)$ denote the probability measure on Ω such that

- (i) $\mathbb{P}_{x,\mu}(X_0 = x) = 1$.
- (ii) For all $f \in C^\infty(M)$,

$$M_t^f = f(X_t) - f(x) - \int_0^t (L_\mu f)(X_s) ds$$

is a $\mathbb{P}_{x,\mu}$ -martingale relative to $\{\mathcal{B}_t: t \geq 0\}$.

In other words, $\mathbb{P}_{x,\mu}$ is the law of the diffusion process $\{Y_t\}$ with initial condition x and generator L_μ solution to the SDE:

$$dY_t = \sum_{i=1}^N F_i(Y_t) \circ dB_t^i - \nabla V_\mu(Y_t) dt, \quad Y_0 = x. \tag{11}$$

In the following $E_{x,r,\mu}$ and $E_{x,\mu}$ will respectively denote the expectation with respect to $\mathbb{P}_{x,r,\mu}$ and to $\mathbb{P}_{x,\mu}$.

3. The Girsanov transform and some lemmas

3.1. The Girsanov transform

Let $B_t = (B_t^1, \dots, B_t^N)$ be a standard Brownian motion on \mathbb{R}^N , \mathbb{P} the law of $(B_s; s \geq 0)$, \mathbb{E} the associated expectation, \mathcal{F}_t the \mathbb{P} -completion of $\sigma(B_s, 0 \leq s \leq t)$ and $\mathcal{F} = \mathcal{F}_\infty$. Let $\{W_t^x\}$ be the solution to the SDE

$$dW_t^x = \sum_{i=1}^N F_i(W_t^x) \circ dB_t^i, \quad W_0^x = x \in M. \tag{12}$$

Then $W^x = (W_t^x, t \geq 0)$ is a Brownian motion on M starting at x . We denote its law \mathbb{P}_x . Note that $W^x : C(\mathbb{R}^+; \mathbb{R}^N) \rightarrow \Omega = C(\mathbb{R}^+; M)$ is measurable. Let

$$M_t^{x,r,\mu} = \exp \left[\int_0^t \sum_i \langle \nabla V_{\mu_s^x(r,\mu)}(W_s^x), F_i(W_s^x) \rangle dB_s^i - \frac{1}{2} \int_0^t \|\nabla V_{\mu_s^x(r,\mu)}(W_s^x)\|^2 ds \right], \tag{13}$$

where

$$\mu_t^x(r, \mu) = \frac{1}{t+r} \left(r\mu + \int_0^t \delta_{W_s^x} ds \right), \tag{14}$$

and

$$M_t^{x,\mu} = \exp \left[\int_0^t \sum_i \langle \nabla V_\mu(W_s^x), F_i(W_s^x) \rangle dB_s^i - \frac{1}{2} \int_0^t \|\nabla V_\mu(W_s^x)\|^2 ds \right]. \tag{15}$$

Then $\{M_t^{x,r,\mu}\}$ and $\{M_t^{x,\mu}\}$ are $(\mathbb{P}, \{\mathcal{F}_t\})$ -martingales. By the transformation of drift formula (see [3], Section IV 4.1 and Theorem IV 4.2),

$$\begin{cases} \mathbb{E}_{x,r,\mu}[Z_t] = \mathbb{E}[M_t^{x,r,\mu}(Z_t \circ W^x)], \\ \mathbb{E}_{x,\mu}[Z_t] = \mathbb{E}[M_t^{x,\mu}(Z_t \circ W^x)] \end{cases} \tag{16}$$

for every bounded \mathcal{B}_t -measurable random variable Z_t . Note that this implies in particular that \mathbb{P}_x , $\mathbb{P}_{x,\mu}$ and $\mathbb{P}_{x,r,\mu}$ are equivalent.

3.2. Some lemmas

The next lemma is a basic tool to estimate quantities such as

$$\mathbb{E}_{x,r,\mu^r}[Z_t] - \mathbb{E}_{x,\mu}[Z_t],$$

for large r and μ^r close to μ .

LEMMA 3.1. – For $a = 1, 2$ let $A_t^a = (A_t^{a,1}, \dots, A_t^{a,N})$ be a \mathbb{R}^N -valued bounded $\{\mathcal{F}_t\}$ -previsible process. Suppose that for all $0 \leq s \leq t$

$$\|A_s^1 - A_s^2\| \leq \delta(t) \tag{17}$$

for some deterministic function $\delta : \mathbb{R}_+ \rightarrow \mathbb{R}_+$. Let

$$M_t^a = \exp \left[\int_0^t \sum_i A_s^{a,i} dB_s^i - \frac{1}{2} \int_0^t \|A_s^a\|^2 ds \right], \quad a = 1, 2.$$

Then there exists a positive constant C such that for any \mathcal{F}_t -measurable random variable Z_t bounded by 1 in absolute value,

$$|\mathbb{E}[M_t^1 Z_t] - \mathbb{E}[M_t^2 Z_t]| \leq e^{Ct} \delta(t). \tag{18}$$

The constant C depends only on $\sup_{a,s} \|A_s^a\|_\infty$.

Lemma 3.1 will be proved in Section 8. Note that Lemma 3.1 and the Girsanov transforms given in Section 3.1 imply that $\mathbb{P}_{x,r,\mu}$ converges weakly towards $\mathbb{P}_{x,\mu}$ as $r \rightarrow \infty$. More precisely

LEMMA 3.2. – *There exists a positive constant C (depending only on $\sup_{x,y} \|\nabla V_y(x)\|$) such for any \mathcal{B}_t -measurable random variable Z_t bounded by 1 in absolute value,*

$$|\mathbb{E}_{x,r,\mu}[Z_t] - \mathbb{E}_{x,\mu}[Z_t]| \leq \frac{e^{Ct}}{r+t}. \tag{19}$$

Proof. – There exists a constant C such that $\|\nabla V_\mu\|_\infty \leq C$ and $\|\nabla V_{\mu_s(r,\mu)} - \nabla V_\mu\|_\infty \leq Ct/(r+t)$, for all $0 \leq s \leq t$. The result then follows from Girsanov formulas (16) and Lemma 3.1 applied with

$$\begin{cases} A_s^{1,i} = \langle \nabla V_{\mu_s(r,\mu)}(W_s^x), F_i(W_s^x) \rangle, \\ A_s^{2,i} = \langle \nabla V_\mu(W_s^x), F_i(W_s^x) \rangle. \end{cases} \quad \square \tag{20}$$

4. The asymptotic of $\mathbb{P}_{X_t, r+t, \mu_t(r,\mu)}$

Here we shall prove:

THEOREM 4.1. – *Let $\mu^* : \Omega \rightarrow \mathcal{P}(M)$ denote a $\mathcal{P}(M)$ -valued random variable. Let $\tilde{\Omega} = \{w \in \Omega : \lim_{t \rightarrow \infty} \mu_t(r, \mu, w) = \mu^*\}$. Then $\mathbb{P}_{x,r,\mu}$ -a.s. on $\tilde{\Omega}$,*

$$\lim_{t \rightarrow \infty} d(\mathbb{P}_{X_t, r+t, \mu_t(r,\mu)}, \mathbb{P}_{X_t, \mu^*}) = 0. \tag{21}$$

COROLLARY 4.2. – *For every bounded and continuous function $Z : \Omega \rightarrow \mathbb{R}$, $\mathbb{P}_{x,r,\mu}$ -a.s.*

$$\lim_{r \rightarrow \infty} |\mathbb{E}_{x,r,\mu}[Z \circ \theta_t | \mathcal{B}_t] - \mathbb{E}_{X_t, \mu^*}[Z]| \mathbf{1}_{\tilde{\Omega}} = 0, \tag{22}$$

where $\theta_t : \Omega \rightarrow \Omega$ is the shift on Ω defined by $\theta_t(w)(s) = w(t+s)$.

Proof. – By the Markov property, we have

$$\mathbb{E}_{x,r,\mu}[Z \circ \theta_t | \mathcal{B}_t] = \mathbb{E}_{X_t, t+r, \mu_t(r,\mu)}[Z], \tag{23}$$

and the result follows from Theorem 4.1. \square

Proof of Theorem 4.1. – Follows directly from the following estimate:

PROPOSITION 4.3. – Let $\{\mu^r : r > 0\} \subset \mathcal{P}(M)$. Assume that

$$\lim_{r \rightarrow \infty} \mu^r = \mu^*$$

in $\mathcal{P}(M)$. Let Z_t be a random variable \mathcal{F}_t -measurable and bounded by 1 in absolute value. Then

$$\lim_{r \rightarrow \infty} \mathbb{E}_{x,r,\mu^r}[Z_t] = \mathbb{E}_{x,\mu^*}[Z_t] \tag{24}$$

uniformly in $x \in M$.

More precisely, there exists $C > 0$ (depending only on $\sup_{x,y} \|\nabla V_y(x)\|$) such that

$$|\mathbb{E}_{x,r,\mu^r}[Z_t] - \mathbb{E}_{x,\mu^*}[Z_t]| \leq e^{Ct} \left(\frac{1}{r} + \varepsilon(r) \right), \tag{25}$$

where $\varepsilon(r) = \sup_x \|\nabla V_{\mu^r}(x) - \nabla V_{\mu^*}(x)\|$.

Note that $\lim_{r \rightarrow \infty} \varepsilon(r) = 0$ (since $x \mapsto \nabla V_{\mu^r}(x) - \nabla V_{\mu^*}(x)$ is equicontinuous in x and converges towards 0 for every x).

Proof. – Lemma 3.1 applied with

$$\begin{cases} A_s^{1,i} = \langle \nabla V_{\mu^r}(W_s^x), F_i(W_s^x) \rangle, \\ A_s^{2,i} = \langle \nabla V_{\mu^*}(W_s^x), F_i(W_s^x) \rangle \end{cases} \tag{26}$$

implies

$$|\mathbb{E}_{x,\mu^r}[Z_t] - \mathbb{E}_{x,\mu^*}[Z_t]| \leq e^{Ct} \varepsilon(r). \tag{27}$$

The conclusion follows from this last inequality combined with Lemma 3.2 and the triangle inequality. \square

5. The convergence in law of X_t

For every $\mu \in \mathcal{P}(M)$ let $\Pi(\mu) \in \mathcal{P}(M)$ denote the invariant probability measure of the diffusion process with generator L_μ . That is

$$\Pi(\mu)(dx) = \frac{e^{-2V_\mu(x)}}{Z(\mu)} \lambda(dx), \tag{28}$$

where $Z(\mu)$ is the normalization constant.

Let us first remark that as $r \rightarrow \infty$, the law of X_t under $\mathbb{P}_{x,r,\mu}$ converges weakly towards the law of X_t under $\mathbb{P}_{x,\mu}$ (see Lemma 3.2). We also have the convergence $\lim_{t \rightarrow \infty} \mathbb{E}_{x,\mu}[g(X_t)] = \Pi(\mu)g^1$ for all $g \in C^0(M)$. The next proposition shows that

¹ For a measure μ and $f \in L^1(\mu)$ we let μf denote $\int f d\mu$.

$\mathbb{E}_{x,r,\mu}[g(X_{t+s})|\mathcal{B}_t] = \mathbb{E}_{X_{t,r+t},\mu_t(r,\mu)}[g(X_s)]$ and $\Pi(\mu_t(r,\mu))g$ are close when s and t tends to ∞ at a certain rate.

PROPOSITION 5.1. – For all $t \geq 1$, $r > 0$, $s > 0$ and $g \in C^0(M)$,

$$|\mathbb{E}_{x,r,\mu}[g(X_{t+s})|\mathcal{B}_t] - \Pi(\mu_t)g| \leq \|g\|_\infty \left(\frac{e^{Cs}}{r+s+t} + C e^{-s/\kappa} \right), \quad (29)$$

where C and κ are positive constants depending only on V .

The proof of Proposition 5.1 is given in Section 8.

COROLLARY 5.2. –

(i) For all positive s and all $g \in C^0(M)$,

$$\limsup_{t \rightarrow \infty} |\mathbb{E}_{x,r,\mu}[g(X_{t+s})|\mathcal{B}_t] - \Pi(\mu_t)g| \leq C \|g\|_\infty e^{-s/\kappa}. \quad (30)$$

(ii) Let s be a real valued positive function such that

$$1 \ll \exp(s(t)) \ll t^{1/C} \quad (31)$$

when t tends to ∞ . Then for all $g \in C^0(M)$,

$$\limsup_{t \rightarrow \infty} |\mathbb{E}_{x,r,\mu}[g(X_{t+s(t)})|\mathcal{B}_t] - \Pi(\mu_t)g| = 0. \quad (32)$$

Proof. – Straightforward. \square

Remark 5.3. – Let \mathcal{L}_t denote the law of $X_{t+s(t)}$ knowing \mathcal{B}_t . Then Corollary 5.2 means that \mathcal{L}_t is asymptotically equal to $\Pi(\mu_t)$. That is, $\lim_{t \rightarrow \infty} \text{dist}_w(\mathcal{L}_t, \Pi(\mu_t)) = 0$, where dist_w is a distance on $\mathcal{P}(M)$ for the weak topology.

Remark that Proposition 5.1 and Corollary 5.2 make no assumption on the asymptotic of $\{\mu_t\}$. Let $\tilde{\Omega} \in \mathcal{B}$ be the event that “ μ_t converges towards μ^* ”, where μ^* is a $\mathcal{P}(M)$ -valued random variable. In [1] and [2], several examples of self-interacting diffusions for which $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$ are given (these examples are shortly presented in Section 7). The following theorem describes the law of $X_{t+s(t)}$ knowing \mathcal{B}_t on $\tilde{\Omega}$.

THEOREM 5.4. – Let $s(t)$ be as in Corollary 5.2. Then, the law of $X_{t+s(t)}$ knowing \mathcal{B}_t converges weakly towards μ^* $\mathbb{P}_{x,r,\mu}$ -a.s. on $\tilde{\Omega}$. That is, for all $g \in C^0(M)$,

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu}[g(X_{t+s(t)})|\mathcal{B}_t] = \mu^*g \quad (33)$$

$\mathbb{P}_{x,r,\mu}$ -a.s. on $\tilde{\Omega}$.

Proof. – It follows from Theorem 3.8 in [1] that μ^* is (almost surely on $\tilde{\Omega}$) a fixed point of Π , i.e., $\Pi(\mu^*) = \mu^*$. The proof now follows from Corollary 5.2(ii) and the fact that $\Pi: \mathcal{P}(M) \rightarrow \mathcal{P}(M)$ is continuous. \square

COROLLARY 5.5 (Convergence in law). – For all $g \in C^0(M)$,

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu} [g(X_t) \mathbf{1}_{\tilde{\Omega}}] = \mathbb{E}_{x,r,\mu} [(\mu^* g) \mathbf{1}_{\tilde{\Omega}}]. \tag{34}$$

In particular, if $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$ then for all $g \in C^0(M)$,

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu} [g(X_t)] = \mathbb{E}_{x,r,\mu} [\mu^* g], \tag{35}$$

i.e., X_t converges in law towards $\mathbb{E}_{x,r,\mu}[\mu^*]$ when t tends to ∞ .

Proof. – In view of Theorem 5.4

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu} [\mathbb{E}_{x,r,\mu} [g(X_{t+s(t)}) | \mathcal{B}_t] \mathbf{1}_{\tilde{\Omega}}] = \mathbb{E}_{x,r,\mu} [(\mu^* g) \mathbf{1}_{\tilde{\Omega}}]. \tag{36}$$

It then suffices to prove that $\lim_{t \rightarrow \infty} a_t = 0$ where

$$a_t = \mathbb{E}_{x,r,\mu} [\mathbb{E}_{x,r,\mu} [g(X_{t+s(t)}) | \mathcal{B}_t] \mathbf{1}_{\tilde{\Omega}} - \mathbb{E}_{x,r,\mu} [g(X_{t+s(t)}) \mathbf{1}_{\tilde{\Omega}} | \mathcal{B}_t]]. \tag{37}$$

Let $\Delta_t = \mathbf{1}_{\tilde{\Omega}} - \mathbb{E}_{x,r,\mu}[\mathbf{1}_{\tilde{\Omega}} | \mathcal{B}_t]$. Then

$$a_t = \mathbb{E}_{x,r,\mu} [\mathbb{E}_{x,r,\mu} [g(X_{t+s(t)}) | \mathcal{B}_t] \Delta_t - \mathbb{E}_{x,r,\mu} [g(X_{t+s(t)}) \Delta_t | \mathcal{B}_t]]. \tag{38}$$

Hence $|a_t| \leq 2 \|g\|_{\infty} \mathbb{E}_{x,r,\mu}[|\Delta_t|]$ and consequently $\lim_{t \rightarrow \infty} a_t = 0$ because $\lim_{t \rightarrow \infty} \Delta_t = 0$ a.s. \square

6. The convergence in law of $(X_{t+u}, u \geq 0)$

In the previous section we were only interested by the asymptotic of the law of X_{t+s} knowing \mathcal{B}_t . These results can be extended to the law of $(X_{t+s+u}; u \geq 0)$ knowing \mathcal{B}_t . The following proposition is analogous to Proposition 5.1 (and implies Proposition 5.1).

PROPOSITION 6.1. – For all $t \geq 1, s > 0, u > 0$ and Z_u a \mathcal{B}_u -measurable random variable bounded by 1 in absolute value, then

$$|\mathbb{E}_{x,r,\mu} [Z_u \circ \theta_{t+s} | \mathcal{B}_t] - \mathbb{E}_{\Pi(\mu_t), \mu_t} [Z_u]| \leq \left(\frac{e^{C(s+u)}}{r+s+u+t} + C e^{-s/\kappa} \right), \tag{39}$$

where C and κ are positive constants depending only on V .

The proof of Proposition 6.1 is given in Section 8.

COROLLARY 6.2. – For any positive u and Z_u a \mathcal{B}_u -measurable random variable bounded by 1 in absolute value, we have

(i) For any positive s ,

$$\limsup_{t \rightarrow \infty} |\mathbb{E}_{x,r,\mu} [Z_u \circ \theta_{t+s} | \mathcal{B}_t] - \mathbb{E}_{\Pi(\mu_t), \mu_t} [Z_u]| \leq C e^{-s/\kappa}. \tag{40}$$

(ii) Let s be a function like in Corollary 5.2, then

$$\limsup_{t \rightarrow \infty} |\mathbb{E}_{x,r,\mu}[Z_u \circ \theta_{t+s(t)} | \mathcal{B}_t] - \mathbb{E}_{\Pi(\mu_t), \mu_t}[Z_u]| = 0. \tag{41}$$

Proof. – Straightforward. \square

This corollary shows that the law of $(X_{t+s(t)+v}; v \geq 0)$ knowing \mathcal{B}_t is asymptotically equal to the law of a diffusion with generator L_{μ_t} and initial distribution $\Pi(\mu_t)$. In particular, (ii) says that

$$\lim_{t \rightarrow \infty} d(\mathbb{P}_t, \mathbb{P}_{\Pi(\mu_t), \mu_t}) = 0, \tag{42}$$

where \mathbb{P}_t is the law of $(X_{t+s(t)+u}; u \geq 0)$ knowing \mathcal{B}_t .

Like in the previous section, we now focus on $\tilde{\Omega}$. The following theorem shows that on $\tilde{\Omega}$, given \mathcal{B}_t , $(X_{t+s(t)+u}; u \geq 0)$ converges in law towards a diffusion with generator L_{μ^*} and initial distribution μ^* (note that μ^* satisfies $\mu^* = \Pi(\mu^*)$ so that μ^* is the invariant probability measure of this diffusion).

THEOREM 6.3. – For any positive u and Z_u a bounded \mathcal{B}_u -measurable random variable,

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu}[Z_u \circ \theta_{t+s(t)} | \mathcal{B}_t] = \mathbb{E}_{\mu^*, \mu^*}[Z_u] \tag{43}$$

almost surely on $\tilde{\Omega}$, where $s(t)$ is as in Corollary 5.2.

Proof. – The proof is the same as the one of Theorem 5.4. \square

Note that Theorem 6.3 implies that on $\tilde{\Omega}$, \mathbb{P}_t converges weakly towards $\mathbb{P}_{\mu^*, \mu^*}$.

COROLLARY 6.4 (Convergence in law). – For any positive u and Z_u a bounded \mathcal{B}_u -measurable random variable,

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu}[(Z_u \circ \theta_t) \mathbf{1}_{\tilde{\Omega}}] = \mathbb{E}_{x,r,\mu}[\mathbb{E}_{\mu^*, \mu^*}[Z_u] \mathbf{1}_{\tilde{\Omega}}]. \tag{44}$$

In particular, if $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$ then

$$\lim_{t \rightarrow \infty} \mathbb{E}_{x,r,\mu}[Z_u \circ \theta_t] = \mathbb{E}_{x,r,\mu}[\mathbb{E}_{\mu^*, \mu^*}[Z_u]]. \tag{45}$$

Proof. – The proof is the same as the one of Corollary 5.5. \square

Note that (44) and (45) respectively imply that the law of $(X_{t+u}; u \geq 0)$ given $\tilde{\Omega}$ converges weakly towards $\mathbb{E}_{x,r,\mu}[\mathbb{P}_{\mu^*, \mu^*} | \tilde{\Omega}]$ and that $\mathbb{E}_{x,r,\mu}[\mathbb{P}_{X_t, r+t, \mu_t(r, \mu)}]$ converges weakly towards $\mathbb{E}_{x,r,\mu}[\mathbb{P}_{\mu^*, \mu^*}]$ provided $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$.

7. Examples

Set $\delta_V(x, y) = \sup_{u \in M} (V_u(x) - V_u(y)) - \inf_{u \in M} (V_u(x) - V_u(y))$. In [1], Corollary 4.4, it is proved that when $\sup_{(x,y) \in M^2} \delta_V(x, y) < 1$, then Π has a unique fixed point μ^*

and $\lim_{t \rightarrow \infty} \mu_t(r, \mu) = \mu^*$ $\mathbb{P}_{x,r,\mu}$ -a.s. The associated self-interacting diffusions produce examples for which $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$, but the limit μ^* is not random.

From the different interactions, we distinguish those such that V is symmetric and defines a positive or a negative self-adjoint operator acting on $L^2(\lambda)$, that can be written in the form $V = \alpha \int_C G(u, x)G(u, y)v(du)$, where C is compact, v is a Borel probability measure, $G : C \times M \rightarrow \mathbb{R}$ is continuous and $\alpha \in \mathbb{R}$. We call them gradient interactions. These interactions produce examples for which $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 1$ and the limit μ^* may be random (see [2]).

When α is positive, we say it is a self-repelling interaction and when α is negative, we say it is a self-attracting interaction. It can be proved (see [2]) that, if $V1$ is a constant function, for all repelling cases or weakly attracting cases ($\alpha > -\alpha_G$, with $\alpha_G > 0$), the empirical occupation measure of the associated self-interacting diffusion converges towards λ a.s. But, when $\alpha < -\alpha_G$, this is not the case, and μ_t may converge towards $\mu^* \neq \lambda$.

The interaction, on the n -dimensional sphere \mathbf{S}^n ,

$$V(x, y) = 2\alpha \cos(d(x, y)) \tag{46}$$

is a gradient interaction. This example is developed in [1], Section 4.2. When $\alpha \geq -(n + 1)/4$, μ_t converges towards λ a.s. and when $\alpha < -(n + 1)/4$, there exists a \mathbf{S}^n -valued random variable v such that μ_t converges a.s. towards $\exp[\beta_n(\alpha) \cos(d(x, v))]\lambda(dx)/Z_{n,\alpha}$, where $Z_{n,\alpha}$ is the normalization constant and $\beta_n(\alpha)$ is a constant depending only on n and α . In [1], Section 4.2, an example of interaction on \mathbf{S}^n (which is not a gradient interaction) for which $\mathbb{P}_{x,r,\mu}(\tilde{\Omega}) = 0$ is given.

8. Proofs

8.1. Proof of Lemma 3.1

Let C be a constant such that both $\|A_t^a\|^2$ and $\|A_t^a\|$ are lower than C . Let

$$E_t = \exp \left[\int_0^t \langle A_s^1, A_s^1 - A_s^2 \rangle ds \right], \tag{47}$$

and $N_t = M_t^2(M_t^1 E_t)^{-1}$. Observe that M_t^a and N_t are exponential martingales solutions of the SDEs

$$\begin{cases} dM_t^a = M_t^a \left(\sum_i A_t^{a,i} dB_t^i \right), \\ dN_t = N_t \left(\sum_i (A_t^{2,i} - A_t^{1,i}) dB_t^i \right). \end{cases} \tag{48}$$

Therefore

$$\begin{cases} \frac{d}{ds} \mathbb{E}[(M_s^a)^2] = \mathbb{E}[(M_s^a)^2 \|A_s^a\|^2] \leq C \mathbb{E}[(M_s^a)^2], \\ \frac{d}{ds} \mathbb{E}[(N_s)^2] = \mathbb{E}[(N_s)^2 \|A_s^1 - A_s^2\|^2] \leq \delta^2(t) \mathbb{E}[(N_s)^2], \end{cases} \tag{49}$$

for $s \leq t$. Hence, by Gronwall’s lemma, for $a \in \{1, 2\}$

$$\begin{cases} \mathbb{E}[(M_t^a)^2] \leq e^{Ct}, \\ \mathbb{E}[(N_t)^2] \leq \exp(t\delta^2(t)). \end{cases} \tag{50}$$

Notice that we also have

$$|E_t - 1| \leq \exp(Ct\delta(t)) - 1. \tag{51}$$

Using these estimates and Schwartz inequality, we get

$$\begin{aligned} |\mathbb{E}[M_t^2 Z_t] - \mathbb{E}[M_t^1 Z_t]| &= |\mathbb{E}[Z_t(N_t E_t - 1)M_t^1]| \\ &\leq \mathbb{E}[(N_t(E_t - 1) + N_t - 1)^2]^{1/2} \mathbb{E}[(M_t^1)^2]^{1/2} \\ &\leq e^{Ct/2} \left[(\exp(Ct\delta(t)) - 1) \exp\left(\frac{t\delta^2(t)}{2}\right) + (\exp(t\delta^2(t)) - 1)^{1/2} \right]. \end{aligned}$$

Since $e^u - 1 \leq u e^u$ we easily obtain

$$|\mathbb{E}_{x,r,\mu}[Z_t] - \mathbb{E}_{x,\mu}[Z_t]| \leq e^{Ct} \delta(t), \tag{52}$$

for C large enough. This proves the lemma. \square

8.2. Proof of Propositions 5.1 and 6.1

Let $P^\mu = (P_t^\mu)_{t \geq 0}$ denote the semigroup of the diffusion with generator L_μ .

LEMMA 8.1. – *Let $g : M \rightarrow \mathbb{R}$ be a bounded continuous function, then for $t \geq 1$,*

$$|P_t^\mu g(x) - \Pi(\mu)g| \leq C \|g\|_\infty e^{-t/\kappa}, \tag{53}$$

for some constant C and κ depending only on $\|V\|_\infty$.

Proof. – Let $\|\cdot\|_2$ be the L^2 -norm defined by

$$\|f\|_2^2 = \int_M f^2(x) \Pi(\mu)(dx). \tag{54}$$

Then, by standard semigroup inequalities (see [1], Section 5.2)

$$\|P_t^\mu g - \Pi(\mu)g\|_2 \leq e^{-t/\kappa} \|g - \Pi(\mu)g\|_2, \quad t > 0, \tag{55}$$

$$\|P_t^\mu g - \Pi(\mu)g\|_\infty \leq C t^{-n/2} \|g - \Pi(\mu)g\|_2, \quad 0 < t \leq 1, \tag{56}$$

for some constant $\kappa > 0$ and $0 < C < \infty$ depending only on $\|V\|_\infty$. Combining (55) and (56) leads to

$$\begin{aligned} \|P_s^\mu g - \Pi(\mu)g\|_\infty &= \|P_1^\mu (P_{s-1}^\mu (g - \Pi(\mu)g))\|_\infty \\ &\leq C e^{-(s-1)/\kappa} \|g - \Pi(\mu)g\|_2 \\ &\leq 2C e^{-(s-1)/\kappa} \|g\|_\infty \end{aligned}$$

for all $s > 1$. \square

Proof of Proposition 5.1. – By the Markov property

$$\mathbf{E}_{x,r,\mu} [g(X_{t+s}) | \mathcal{B}_t] = \mathbf{E}_{X_t, r+t, \mu_t(r,\mu)} [g(X_s)]. \tag{57}$$

Hence

$$\begin{aligned} &|\mathbf{E}_{x,r,\mu} [g(X_{t+s}) | \mathcal{B}_t] - \Pi(\mu_t)g| \\ &\leq |\mathbf{E}_{X_t, r+t, \mu_t(r,\mu)} [g(X_s)] - \mathbf{E}_{X_t, \mu_t} [g(X_s)]| + |\mathbf{E}_{X_t, \mu_t} [g(X_s)] - \Pi(\mu_t)g| \end{aligned}$$

and the result follows from Lemmas 3.2 and 8.1. \square

Proof of Proposition 6.1. – This is almost the same proof. By the Markov property

$$\mathbf{E}_{x,r,\mu} [Z_u \circ \theta_{t+s} | \mathcal{B}_t] = \mathbf{E}_{X_t, r+t, \mu_t(r,\mu)} [Z_u \circ \theta_s].$$

Hence

$$\begin{aligned} &|\mathbf{E}_{x,r,\mu} [Z_u \circ \theta_{t+s} | \mathcal{B}_t] - \mathbf{E}_{\Pi(\mu_t), \mu_t} [Z_u]| \\ &\leq |\mathbf{E}_{X_t, r+t, \mu_t(r,\mu)} [Z_u \circ \theta_s] - \mathbf{E}_{X_t, \mu_t} [Z_u \circ \theta_s]| + |\mathbf{E}_{X_t, \mu_t} [Z_u \circ \theta_s] - \mathbf{E}_{\Pi(\mu_t), \mu_t} [Z_u]|. \end{aligned}$$

The first term of the right-hand side of preceding equation can be dominated using Lemma 3.2. For the domination of the second term, let $\varphi(x) = \mathbf{E}_{x, \mu_t} [Z_u]$, then

$$\begin{cases} \mathbf{E}_{X_t, \mu_t} [Z_u \circ \theta_s] = P_s^{\mu_t} \varphi(X_t), \\ \mathbf{E}_{\Pi(\mu_t), \mu_t} [Z_u] = \Pi(\mu_t) \varphi. \end{cases} \tag{58}$$

We then conclude using Lemma 8.1. \square

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