

# Asymptotic behaviour of some self-interacting diffusions

Aline Kurtzmann

joint works with V. Kleptsyn, P. Del Moral and J. Tugaut

Workshop Nonlinear Processes and their Applications (Saint Etienne)

July 03, 2019

# Outline

## 1 Generalities

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system
- 4 Centering

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system
- 4 Centering
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system
- 4 Centering
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)

What is a self-interacting diffusion?

- Solution of

$$dX_t = dB_t - F(t, X_t, \mu_t)dt$$

- $\mu_t = \frac{1}{t} \int_0^t \delta_{X_s} dS$



# Brownian polymer

Durrett and Rogers (1992) on  $\mathbb{R}^d$ :

$$dX_t = dB_t + \int_0^t f(X_t - X_s) ds dt,$$

where  $f : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is measurable and bounded.

# Brownian polymer

Durrett and Rogers (1992) on  $\mathbb{R}^d$ :

$$dX_t = dB_t + \int_0^t f(X_t - X_s) ds dt,$$

where  $f : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is measurable and bounded.

Question: what is the normalisation of  $X$ ?

# Self-attracting case

Studied by:

- Cranston & Le Jan (1995): linear and  $1 - d$  constant interaction  
( $f(x) = -a \operatorname{sign}(x)$ ),

# Self-attracting case

Studied by:

- Cranston & Le Jan (1995): linear and  $1 - d$  constant interaction ( $f(x) = -a \operatorname{sign}(x)$ ),
- Raimond (1997): constant interaction ( $d \geq 2$ ,  $f(x) = -ax/|x|$  with  $a > 0$ ),

# Self-attracting case

Studied by:

- Cranston & Le Jan (1995): linear and  $1 - d$  constant interaction ( $f(x) = -a \operatorname{sign}(x)$ ),
- Raimond (1997): constant interaction ( $d \geq 2$ ,  $f(x) = -ax/|x|$  with  $a > 0$ ),
- Herrmann & Roynette (2003):

Theorem (Herrmann & Roynette, 2003)

1) Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be an odd function, decreasing and bounded.

Suppose that there exists  $C, \rho > 0$  and  $k \in \mathbb{N}^*$  such that

$|f(x)| \geq Ce^{-\rho/|x|^k}$  around 0. Then  $X_t$  converges a.s.

2) When the interaction is not local,  $f(x) = -\operatorname{sign}(x) \mathbf{1}_{\{|x| \geq a\}}$ , then the trajectories remain bounded a.s. (but do not converge).

# Reinforced diffusion on a compact set

Benaïm, Ledoux and Raimond (2002), Benaïm and Raimond (2003, 2005) on a compact manifold:

$$dX_t = dB_t - \frac{1}{t} \int_0^t \nabla_x W(X_t, X_s) ds dt$$

# Reinforced diffusion on a compact set

Benaïm, Ledoux and Raimond (2002), Benaïm and Raimond (2003, 2005) on a compact manifold:

$$dX_t = dB_t - \frac{1}{t} \int_0^t \nabla_x W(X_t, X_s) ds dt$$

Heuristic: show that  $\mu_t := \frac{1}{t} \int_0^t \delta_{X_s} ds$  is close to a deterministic flow.

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)**
- 3 Discretisation and dynamical system
- 4 Centering
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)



## Study

$$\begin{aligned}dX_t &= dB_t - \left( \frac{1}{t} \int_0^t W'(X_t - X_s) ds \right) dt \\ &= dB_t - W' * \mu_t(X_t) dt\end{aligned}$$

$$\mu_t = \frac{1}{t} \int_0^t \delta_{X_s} ds$$

Example: quadratic  $W$ 

## Lemma

*Let  $W(x) = ax^2$  with  $a > 0$ . Then a.s. the empirical measure  $\mu_t$  converges (weakly) to  $\mu_\infty \sim \mathcal{N}(\bar{\mu}_\infty, 1/a)$ .*

# Example: quadratic $W$

## Lemma

*Let  $W(x) = ax^2$  with  $a > 0$ . Then a.s. the empirical measure  $\mu_t$  converges (weakly) to  $\mu_\infty \sim \mathcal{N}(\bar{\mu}_\infty, 1/a)$ .*

*Let  $W(x) = \frac{1}{2}(x - 1)^2$ . Then  $\bar{\mu}_t = \frac{1}{t} \int_0^t X_s ds$  and  $X_t$  diverge a.s.*

# Set of hypotheses on the interaction potential (H)

- $W$  is  $\mathcal{C}^2$ , strictly uniformly convex and symmetric,

# Set of hypotheses on the interaction potential (H)

- $W$  is  $\mathcal{C}^2$ , strictly uniformly convex and symmetric,
- there exist  $C, k > 0$  such that

$$|W(x)| + |W'(x)| + |W''(x)| \leq C(1 + |x|^k).$$

# Results

## Theorem

*Suppose that  $W$  satisfies the assumption **(H)**. Then there exists a unique probability density function  $\rho_\infty$  such that a.s.*

$$\mu_t \rightarrow \rho_\infty(\cdot - c_\infty)dx.$$

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system**
- 4 Centering
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)

# Relation with a Markovian system



# Relation with a Markovian system

$\mu_t$  is asymptotically close to the deterministic dynamical system:

$$\dot{\mu} = \Pi(\mu) - \mu,$$

where  $\Pi(\mu) := \frac{1}{Z(\mu)} e^{-2W_*\mu}$ .

# Strategy of the proof

- Compare on  $[T_n, T_{n+1}]$  the trajectories of

$$dX_t = dB_t - W' * \mu_t(X_t)dt$$

with those of the corresponding process where  $\mu_t$  is replaced by  $\mu_{T_n}$ :

$$dY_t = dB_t - W' * \mu_{T_n}(Y_t)dt$$

# Strategy of the proof

- Compare on  $[T_n, T_{n+1}]$  the trajectories of

$$dX_t = dB_t - W' * \mu_t(X_t)dt$$

with those of the corresponding process where  $\mu_t$  is replaced by  $\mu_{T_n}$ :

$$dY_t = dB_t - W' * \mu_{T_n}(Y_t)dt$$

- Estimate the speed of convergence of the empirical measure of  $Y$  toward the invariant probability measure  $\Pi(\mu_{T_n})$

# Strategy for the approximation by a dynamical system

- Compare the flow obtained by the “Euler method”

$$\mu_{[T_n, T_{n+1}]} = \mu_{T_n} + \frac{\Delta T_n}{T_{n+1}} (\mu_{[T_n, T_{n+1}]} - \mu_{T_n} + \text{error})$$

with the flow

$$\dot{\mu} = \frac{1}{T_n} (\Pi(\mu) - \mu)$$

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system
- 4 Centering**
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)

# Reference point

## Definition

The *center* of the probability measure is the point  $c_\mu$  such that  $W' * \mu(c_\mu) = 0$ .

We define the *centered measure*  $\mu^c$  as

$$\mu^c(A) = \mu(A + c_\mu).$$

# The deterministic system

- Comparison with the Ornstein-Uhlenbeck process

# The deterministic system

- Comparison with the Ornstein-Uhlenbeck process
- Lyapunov function: free energy



# The deterministic system

- Comparison with the Ornstein-Uhlenbeck process
- Lyapunov function: free energy
- Estimation of the speed of convergence (decrease of the entropy)

# The deterministic system

- Comparison with the Ornstein-Uhlenbeck process
- Lyapunov function: free energy
- Estimation of the speed of convergence (decrease of the entropy)
- Convergence of the center

# Final result

Theorem (Kleptsyn-K., EJP 2012)

Suppose that  $W$  satisfies the hypothesis **(H)**. Then:

- 1 there exists a unique probability density function  $\rho_\infty$  *centered* such that a.s.

$$\mu_t^c \rightarrow \rho_\infty(x)dx,$$

# Final result

Theorem (Kleptsyn-K., EJP 2012)

Suppose that  $W$  satisfies the hypothesis **(H)**. Then:

- 1 there exists a unique probability density function  $\rho_\infty$  centered such that a.s.

$$\mu_t^c \rightarrow \rho_\infty(x)dx,$$

- 2 a.s. the center  $c_t = c(\mu_t)$  converges to a (random) limit  $c_\infty$ .

And there exists  $a > 0$  such that a.s., we have for  $t$  large enough

$$\mathbb{W}_2(\mu_t^c, \rho_\infty) = O(\exp\{-a^{2k+1}\sqrt{\log t}\}).$$

# Outline

- 1 Generalities
- 2 Self-attracting diffusion on  $\mathbb{R}$  (with Victor Kleptsyn)
- 3 Discretisation and dynamical system
- 4 Centering
- 5 Kramers' type law (with Pierre Del Moral and Julian Tugaut)

# The considered self-interacting diffusion

$$\begin{aligned} dX_t &= \sigma dB_t - \left( V'(X_t) + \frac{1}{t} \int_0^t W'(X_t - X_s) ds \right) dt \\ &= \sigma dB_t - (V'(X_t) + W' * \mu_t(X_t)) dt \end{aligned}$$

$$\mu_t = \frac{1}{t} \int_0^t \delta_{X_s} ds$$

$$\sigma > 0.$$

# Final result

Theorem (Del Moral- K. -Tugaut, submitted 2019 )

Suppose that  $V$  and  $W$  satisfy the hypothese **(H)**. Let  $m$  be the *unique* minimum of  $V$ .

Let  $\mathcal{D}$  be a stable domain for the flow  $x \mapsto -V'(x) - W'(x - m)$ .

Denote by  $\tau$  the *first time the process  $X$  exits the domain  $\mathcal{D}$* .

Let  $H = \inf_{x \in \partial \mathcal{D}} V(x) + W(x - m) - V(m)$  be the *exit cost* from  $\mathcal{D}$ .

# Final result

Theorem (Del Moral- K. -Tugaut, submitted 2019 )

Suppose that  $V$  and  $W$  satisfy the hypothesis **(H)**. Let  $m$  be the *unique* minimum of  $V$ .

Let  $\mathcal{D}$  be a stable domain for the flow  $x \mapsto -V'(x) - W'(x - m)$ .

Denote by  $\tau$  the *first time the process  $X$  exits the domain  $\mathcal{D}$* .

Let  $H = \inf_{x \in \partial \mathcal{D}} V(x) + W(x - m) - V(m)$  be the *exit cost* from  $\mathcal{D}$ .

Then we have for any  $\delta > 0$

$$\lim_{\sigma \rightarrow 0} \mathbb{P}(e^{2(H-\delta)/\sigma^2} \leq \tau \leq e^{2(H+\delta)/\sigma^2}) = 1.$$



# Strategy of the proof

- Use the previous speed of convergence

# Strategy of the proof

- Use the previous speed of convergence
- $X_t$  and  $\psi_t$  (solution to the ODE  $\dot{\psi}_t = -V'(\psi_t) - \frac{1}{t} \int_0^t W'(\psi_t - \psi_s) ds$ ) are uniformly close as  $\sigma \rightarrow 0$

# Strategy of the proof

- Use the previous speed of convergence
- $X_t$  and  $\psi_t$  (solution to the ODE  $\dot{\psi}_t = -V'(\psi_t) - \frac{1}{t} \int_0^t W'(\psi_t - \psi_s) ds$ ) are uniformly close as  $\sigma \rightarrow 0$
- Probability of leaving a stable domain before the empirical measure remains stuck in  $\mathbb{B}(\delta_m, \kappa)$  goes to zero with  $\sigma$

# Strategy of the proof

- Use the previous speed of convergence
- $X_t$  and  $\psi_t$  (solution to the ODE  $\dot{\psi}_t = -V'(\psi_t) - \frac{1}{t} \int_0^t W'(\psi_t - \psi_s) ds$ ) are uniformly close as  $\sigma \rightarrow 0$
- Probability of leaving a stable domain before the empirical measure remains stuck in  $\mathbb{B}(\delta_m, \kappa)$  goes to zero with  $\sigma$
- Coupling between  $X$  and a Markov diffusion to use former Tugaut's results