

Vanishing noise limit of the first collision and the first collision-location between two self-stabilizing diffusions

a.k.a. A Kramers' type law for the collision time of two self-interacting diffusion processes and of their
related particle approximation

Jean-François Jabir - HSE Moscow & LSA

Joint work with Julian Tugaut, Univ. Jean-Monnet.

Workshop *Metastability, Mean-Field Particles and Non-Linear Processes*,

St Etienne, May 17-20, 2022

Aims: Given $\sigma, \alpha > 0$ and the two independent self-stabilizing diffusion processes:

$$X_t = x_1 + \sigma B_t - \int_0^t \left\{ \nabla V(X_s) + \alpha(X_s - \mathbb{E}[X_s]) \right\} ds, \quad t \geq 0,$$

$$Y_t = x_2 + \sigma \tilde{B}_t - \int_0^t \left\{ \nabla V(Y_s) + \alpha(Y_s - \mathbb{E}[Y_s]) \right\} ds, \quad t \geq 0,$$

establish the $\sigma \downarrow 0$ -asymptotic of

- the first collision-time between X and Y , "given" by $C(\sigma) = \inf\{t \geq 0 : X_t = Y_t\}$,
- the first collision-location $X_{C(\sigma)} (= Y_{C(\sigma)})$.

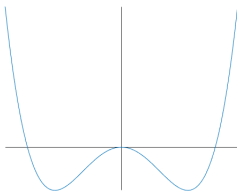
Same questions for the related particle systems:

$$X_t^{i,N} = x_1 + \sigma B_t^i - \int_0^t \left\{ \nabla V(X_s^{i,N}) + \alpha \left(X_s - \frac{1}{N} \sum_{j=1}^N X_s^{j,N} \right) \right\} ds, \quad t \geq 0, \quad 1 \leq i \leq N,$$

$$Y_t^{i,N} = x_2 + \sigma \tilde{B}_t^i - \int_0^t \left\{ \nabla V(Y_s^{i,N}) + \alpha \left(Y_s - \frac{1}{N} \sum_{j=1}^N Y_s^{j,N} \right) \right\} ds, \quad t \geq 0, \quad 1 \leq i \leq N.$$

Prototypical case

Double wells landscape (Kramers '40, Dawson '83): $d = 1$, $V(x) = \frac{x^4}{4} - \frac{x^2}{2}$.



At the limit $\sigma = 0$,

$$X_t \longrightarrow \phi(t, x_1) = x_1 - \int_0^t V'(\phi(s, x_1)) ds,$$

$$Y_t \longrightarrow \phi(t, x_2) = x_2 - \int_0^t V'(\phi(s, x_2)) ds.$$

If $x_1 < -1$ and $1 < x_2$, at $\sigma = 0$ then collision(s) between X and Y can happen due to the action of the Brownian motions.

Heuristic: As $\sigma \downarrow 0$, it is expected that $C(\sigma)$ grows to ∞ at a certain rate, and $X_{C(\sigma)}$ should "persist" in a certain point between the wells $\lambda_1 = -1$ and $\lambda_2 = 1$.

General assumptions

(A) – (0) B and \tilde{B} are two independent \mathbb{R}^d -Brownian motions;

(A) – (i) $V : \mathbb{R}^d \rightarrow \mathbb{R}$ is of class \mathcal{C}^2 , locally Lipschitz, convex at infinity (namely $\inf_{|x| \geq R'} \nabla^2 V(x)$ is positive definite for some $R' > 0$) and is such that ∇V grows at a $2n$ -polynomial rate on \mathbb{R}^d :

$$\sup_{x \in \mathbb{R}^d} \{ (1 + |x|^{2n})^{-1} |\nabla V(x)| \} < \infty.$$

(A) – (ii) V admits two distinct (strict) local minima located in λ_1 and λ_2 .

(A) – (iii) Synchronization condition: α is large enough so that $\alpha I_d + \nabla^2 V$ positive definite.

(A) – (iv) The starting points x_1 and x_2 lie in a respective basin of attraction of V :

$$\phi(t, x_i) = x_i - \int_0^t \nabla V(\phi(s, x_i)) ds \xrightarrow[t \rightarrow \infty]{} \lambda_i, \quad i = 1, 2.$$

Notes:

- (A) – (iii) $\Rightarrow x \mapsto V(x) + \frac{\alpha}{2} |x - m|^2$ is strictly convex, for any $m \in \mathbb{R}^d$.
- (A) – (iv) \Rightarrow No collision at $\sigma = 0$.

Stochastic Cucker-Smale model:

$$\begin{cases} dX_t^{i,N} = U_t^{i,N} dt, \\ dU_t^{i,N} = \sum_{j=1}^N \xi_{i,j}^N(t, X_t^{1,N}, \dots, X_t^{N,N}) (U_t^{j,N} - U_t^{i,N}) dt + \sigma(X_t^{i,N}, U_t^{i,N}) dW_t^i. \end{cases}$$

- $\sigma = 0$: Cucker-Smale model for flocking ([CK07]): Under certain conditions on the interaction/communication rate $\xi_{i,j}^N$, emergence of a common behaviour at large-time:

$$\lim_{t \rightarrow \infty} \max_{i,j} |U_t^{i,N} - U_t^{j,N}| = 0, \quad \lim_{t \rightarrow \infty} \max_{i,j} |X_t^{i,N} - X_t^{j,N}| < \infty.$$

- $\sigma \neq 0$: [CDP18]: Different types of noise (additive, multiplicative, idiosyncratic, common, ...) leads to different probabilistic interpretation of the flocking ($L^p(\Omega)$, a.s., in proba., weak, ...)

Cucker-Smale models with "emerging" leaders and noise-induced collision: Two groups of populations each flocking as $t \rightarrow \infty$ around a prescribed attractor. Initially, no communication/interaction between the two populations, but a random perturbation forces the two populations to interact with each other. As the effect of this perturbation vanishes, what happens of the collision point ?

Self-stabilizing diffusions : Toy models where self-interacting diffusions represents overdamped/Kramers-Smoluchowski limit to the Langevin type models above, where possible leaders are static and possible post-collision effects are neglected.

One-dimensional case versus multidimensional case

- For $d = 1$, true collision happen and

$$C(\sigma) = \inf\{t \geq 0 : X_t = Y_t\}$$

is finite a.s.

- Whenever $d > 1$, B and \tilde{B} do not necessarily hit each others, and without further conditions on V , $C(\sigma)$ is not well-defined \Rightarrow Require an enlargement of the collision.

Enlarged first collision-time/enlarged first collision-location:

$$C_\varepsilon(\sigma) = \inf\{t \geq 0 : |X_t - Y_t| \leq 2\varepsilon\}, \quad (X_{C_\varepsilon(\sigma)}, Y_{C_\varepsilon(\sigma)}).$$

Threshold:

$$\epsilon < \varepsilon_0 := 2^{-1} \inf_{t \geq 0} |\phi(t, x_1) - \phi(t, x_2)|$$

In particular, $\epsilon < \inf\{|x_1 - x_2|, |\lambda_1 - \lambda_2|\}$

Collision-time as an exit-time

$$C_\epsilon(\sigma) = \inf\{t \geq 0 : (X_t, Y_t) \notin (\mathbb{R}^d \times \mathbb{R}^d) \setminus \Delta_\epsilon\},$$

where

$$\Delta_\epsilon := \{(x, y) \in \mathbb{R}^d \times \mathbb{R}^d : |x - y| = 2\epsilon\}.$$

Gradient flow:

$$\Psi_t(z_0) = z_0 - \int_0^t \nabla U(\Psi_s(z_0)) ds, t \geq 0,$$

Perturbed version:

$$z_t^\sigma = z_0 + \sigma \mathcal{B}_t - \int_0^t \nabla U(z_s^\sigma) ds, t \geq 0.$$

Large Deviation Principle: For any finite time horizon T , and for any $\delta > 0$,

$$\lim_{\sigma \rightarrow 0} \frac{\sigma^2}{2} \log \mathbb{P} \left\{ \sup_{t \in [0; T]} |z_t^\sigma - \Psi_t(z_0)| > \delta \right\} = - \inf_{\Phi} \int_0^T \left| \frac{d\Phi}{dt} + \nabla U(\Phi(t)) \right|^2 dt,$$

where the supremum is taken over all possible \mathcal{C}^1 -path Φ starting from z_0 and such that $\max_{0 \leq t \leq T} |\Phi(t) - \Psi_t(z_0)| > \delta$.

Definition

We say that the domain $\mathcal{G} \subset \mathbb{R}^d$ is stable/positively invariant by $-\nabla U$ if for all $z_0 \in \mathcal{G}$ the orbit $\{\Psi_t(z_0); t \in \mathbb{R}_+\}$ remains in \mathcal{G} .

Theorem

Let U be a convex C^2 function and \mathcal{G} be an open bounded set of \mathbb{R}^d , stable by ∇U . Assume also that for all $x_0 \in \partial\mathcal{G}$, $\Psi_t(x_0)$ converges to a unique point a_0 at large time. Then, for any z_0 in \mathcal{G} , and for

$$\tau_{\mathcal{G}}(\sigma) := \inf\{t \geq 0 : z_t^\sigma \notin \mathcal{G}\},$$

we have:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H} - \delta) \right] < \tau_{\mathcal{G}}(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H} + \delta) \right] \right\} = 1,$$

where \underline{H} is the exit-cost of \mathcal{G} :

$$\underline{H} = \inf_{z \in \partial\mathcal{G}} \inf_{T > 0} \inf_{\Phi} \{I_{T,z}(\Phi)\} = \inf_{x \in \partial\mathcal{G}} (U(x) - U(a_0))$$

Additionally, if $\inf_{x \in \partial\mathcal{G}} (U(x) - U(a_0))$ is achieved in a unique point z_\star in $\partial\mathcal{G}$, then, for all $\delta > 0$, $z_0 \in \mathcal{G}$,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ |z_{\tau_{\mathcal{G}}(\sigma)}^\sigma - z_\star| < \delta \right\} = 1.$$

- Hermann, Imkeller and Peithmann [HIP08]: Kramers' type law for the self-stabilizing diffusion:

$$Z_t = y + \sigma B_t - \int_0^t \left\{ \nabla U(Z_s) + \int \nabla \phi(Z_s - y) \mu_s(dy) \right\} ds, \quad \mu_t = \text{Law}(Z_t), \quad t \geq 0.$$

Assuming U and ϕ are relatively smooth and strictly convex functions, and \mathcal{G} is a stable set,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H} - \delta) \right] < \tau_{\mathcal{G}}(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H} + \delta) \right] \right\} = 1,$$

for

$$\underline{H} = \inf_{z \in \partial G} H,$$

$$H(z) = (U(z) + \phi(z - z_*) - U(z_*)), \quad z_* = \operatorname{argmin} U.$$

- Tugaut *et al.* 2007-2021: Kramers' type law in the case of a double wells landscape and other globally non-convex situations (e.g. [T21] for the case of the granular media equation).

Heuristic Kramers' type law for the first collision time between X and Y : If $(\mathbb{R}^d \times \mathbb{R}^d) \setminus \Delta_\varepsilon$ was stable,

$$\log(C_\varepsilon(\sigma)) \approx \frac{2}{\sigma^2} \underline{\mathcal{H}}_\varepsilon, \quad (X_{C_\varepsilon(\sigma)}, Y_{C_\varepsilon(\sigma)}) \approx \mathcal{M}_\varepsilon,$$

for

$$\underline{\mathcal{H}}_\varepsilon = \inf_{(x,y) \in \partial \Delta_\varepsilon} H(x,y), \quad \mathcal{M}_\varepsilon = \operatorname{argmin}_{\partial \Delta_\varepsilon} H,$$

$$H(x,y) = V(x) - V(\lambda_1) + \frac{\alpha}{2}|x - \lambda_1|^2 + V(y) - V(\lambda_2) + \frac{\alpha}{2}|y - \lambda_2|^2.$$

For $\varepsilon \ll 1$, $\sigma \downarrow 0$,

$$\log(C_\varepsilon(\sigma)) \approx \frac{2}{\sigma^2} \inf_x H(x,x), \quad (X_{C_\varepsilon(\sigma)}, Y_{C_\varepsilon(\sigma)}) \approx \operatorname{argmin}_x H(x,x).$$

Heuristic Kramers' type law for the first collision time between X and Y : If $(\mathbb{R}^d \times \mathbb{R}^d) \setminus \Delta_\varepsilon$ was stable,

$$\log(C_\varepsilon(\sigma)) \approx \frac{2}{\sigma^2} \underline{\mathcal{H}}_\varepsilon, \quad (X_{C_\varepsilon(\sigma)}, Y_{C_\varepsilon(\sigma)}) \approx \mathcal{M}_\varepsilon,$$

for

$$\underline{\mathcal{H}}_\varepsilon = \inf_{(x,y) \in \partial \Delta_\varepsilon} H(x,y), \quad \mathcal{M}_\varepsilon = \operatorname{argmin}_{\partial \Delta_\varepsilon} H,$$

$$H(x,y) = V(x) - V(\lambda_1) + \frac{\alpha}{2}|x - \lambda_1|^2 + V(y) - V(\lambda_2) + \frac{\alpha}{2}|y - \lambda_2|^2.$$

For $\varepsilon \ll 1$, $\sigma \downarrow 0$,

$$\log(C_\varepsilon(\sigma)) \approx \frac{2}{\sigma^2} \inf_x H(x,x), \quad (X_{C_\varepsilon(\sigma)}, Y_{C_\varepsilon(\sigma)}) \approx \operatorname{argmin}_x H(x,x).$$

Problem: No *stability* property of $(\mathbb{R}^d \times \mathbb{R}^d) \setminus \Delta_\varepsilon$.

Alternative:

$$\begin{aligned} C_\varepsilon(\sigma) &= \inf_{\lambda \in \mathbb{R}^d} \beta_{\lambda, \varepsilon}(\sigma), \\ \beta_{\lambda, \varepsilon}(\sigma) &:= \inf \{t \geq 0 : (X_t, Y_t) \in B(\lambda, \varepsilon) \times B(\lambda, \varepsilon)\}. \end{aligned}$$

Strategy: (Coupling techniques *a la* J. Tugaut) 1/ Establish preliminaries Kramers' type laws on

$$x_t^\sigma = x_1 + \sigma B_t - \int_0^t \left\{ \nabla V(x_s^\sigma) + \alpha(x_s^\sigma - \lambda_1) \right\} ds, \quad t \geq 0,$$

$$y_t^\sigma = x_2 + \sigma \tilde{B}_t - \int_0^t \left\{ \nabla V(y_s^\sigma) + \alpha(y_s^\sigma - \lambda_2) \right\} ds, \quad t \geq 0.$$

2/ Transfer the Kramers' type laws to the self-stabilizing diffusions.

Dynamics:

$$x_t^\sigma = x_1 + \sigma B_t - \int_0^t \nabla \Psi_1(x_s^\sigma) ds, \quad t \geq 0,$$

$$y_t^\sigma = x_2 + \sigma \tilde{B}_t - \int_0^t \nabla \Psi_2(y_s^\sigma) ds, \quad t \geq 0.$$

Assumptions:

- Ψ_1 and Ψ_2 are of class \mathcal{C}^2 , strictly convex and admits λ_1 and λ_2 as their respective minimizers.
- $\inf_{t \geq 0} |\phi_t^1(x_1) - \phi_t^2(x_2)| =: 2\epsilon_0 > 0$ for $\phi_t^i(z) = z - \int_0^t \nabla \Psi_i(\phi_s^i(z)) ds, \quad t \geq 0, \quad i = 1, 2.$

Approximated first collision time:

$$c_\epsilon(\sigma) := \inf \{t \geq 0 : |x_t^\sigma - y_t^\sigma| \leq 2\epsilon\}, \quad \epsilon < \epsilon_0.$$

Reformulation: $c_\epsilon(\sigma) = \inf_\lambda \beta_{\lambda, \epsilon}(\sigma),$

$$\beta_{\lambda, \rho}(\sigma) = \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \notin (\mathbb{R}^d \times \mathbb{R}^d) \setminus (B(\lambda, \epsilon) \times B(\lambda, \epsilon))\}$$

Additional enlargement: $\widehat{\beta}_{\lambda, \rho}(\sigma) = \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \in \mathcal{D}_{\lambda, \epsilon}^1 \times \mathcal{D}_{\lambda, \epsilon}^2\},$

$$\mathcal{D}_{\lambda, \epsilon}^i := \{\phi^{i,+}(t, x_i) : t \geq 0, z \in B(\lambda, \rho)\}, \quad \phi^{i,+}(t, x_i) = z + \int_0^t \nabla \Psi_i(\phi^{i,+}(s, x_i)) ds, \quad t \geq 0.$$

Advantages:

- Whenever the "prescribed" collision λ is far from the wells (i.e. $\min_{i=1,2}(|\lambda - \lambda_i|) > \epsilon$), the domain $(\mathbb{R}^d \setminus \mathcal{D}_{\lambda,\epsilon}^1) \times (\mathbb{R}^d \setminus \mathcal{D}_{\lambda,\epsilon}^2)$ is stable by $(-\nabla\Psi^1, -\nabla\Psi^2)$.
- Whenever λ is too close to one of the wells, say $|\lambda - \lambda_1| < \epsilon$ then $B(\lambda, \rho)$ is an attractive set for $\phi^{1,+}$ and $\mathcal{D}_{\lambda,\rho}^1 = \mathbb{R}^d$, meanwhile $\mathbb{R}^d \setminus \mathcal{D}_{\lambda,\epsilon}^2$ is stable by $-\nabla\Psi^2$.
- The case $\{|\lambda - \lambda_1| = \epsilon\} \cup \{|\lambda - \lambda_2| = \epsilon\}$ is singular and requires a slight rescaling.

Last approximation: For $0 < \rho < 1$,

$$\widehat{\beta}_{\lambda,\epsilon}^\rho(\sigma) := \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \in \mathcal{O}_{\lambda,\epsilon,\rho}\} \quad (1)$$

where the domain $\mathcal{O}_{\lambda,\epsilon,\rho}$ is given by

$$\mathcal{O}_{\lambda,\epsilon,\rho} := \begin{cases} \mathcal{D}^1(\lambda, \rho\epsilon) \times \mathcal{D}^2(\lambda, \epsilon) & \text{if } |\lambda - \lambda_1| = \epsilon, \\ \mathcal{D}^1(\lambda, \epsilon) \times \mathcal{D}^2(\lambda, \rho\epsilon) & \text{if } |\lambda - \lambda_2| = \epsilon, \\ \mathcal{D}^1(\lambda, \epsilon) \times \mathcal{D}^2(\lambda, \epsilon) & \text{otherwise.} \end{cases}$$

- Applying Freidlin-Wentzell's exit-time estimates:

Lemma

For any λ in \mathbb{R}^d and for any $\delta > 0$,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} \left(\widehat{h}_\epsilon^\rho(\lambda) - \delta \right) \right] < \widehat{\beta}_{\lambda, \epsilon}^\rho(\sigma) < \exp \left[\frac{2}{\sigma^2} \left(\widehat{h}_\epsilon^\rho(\lambda) + \delta \right) \right] \right\} = 1,$$

for

$$\widehat{h}_\epsilon^\rho(\lambda) = \begin{cases} \inf_{x \in \partial B(\lambda; \rho\epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda; \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } |\lambda - \lambda_1| = \epsilon, \\ \inf_{x \in \partial B(\lambda; \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda; \rho\epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } |\lambda - \lambda_2| = \epsilon, \\ \inf_{x \in \mathcal{D}_{\lambda, \epsilon}^1} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \mathcal{D}_{\lambda, \epsilon}^2} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{otherwise,} \end{cases}$$

Moreover, we have: for any $\delta > 0$,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \min \left(\text{dist} \left(x_{\widehat{\beta}_{\lambda, \epsilon}^\rho(\sigma)}^\sigma, B(\lambda, \epsilon) \right), \text{dist} \left(y_{\widehat{\beta}_{\lambda, \epsilon}^\rho(\sigma)}^\sigma, B(\lambda, \epsilon) \right) \right) \leq \delta \right\} = 1,$$

for $\text{dist}(x, B(\lambda, \epsilon)) := \inf_{z \in B(\lambda, \epsilon)} |x - z|$.

- From the domains $\mathcal{D}_{\lambda,\epsilon}^1$ and $\mathcal{D}_{\lambda,\epsilon}^2$ to $B(\lambda, \epsilon)$:

Lemma

The same Kramers' type law holds for

$$\beta_{\lambda,\epsilon}^\rho(\sigma) = \begin{cases} \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \in B(\lambda, \rho\epsilon) \times B(\lambda, \epsilon)\} & \text{if } |\lambda - \lambda_1| = \epsilon, \\ \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \in B(\lambda, \epsilon) \times B(\lambda, \rho\epsilon)\} & \text{if } |\lambda - \lambda_2| = \epsilon, \\ \inf \{t \geq 0 : (x_t^\sigma, y_t^\sigma) \in B(\lambda, \epsilon) \times B(\lambda, \epsilon)\} & \text{otherwise.} \end{cases}$$

- Asymptotic $\rho = 1$: $\widehat{h}_\epsilon^1(\lambda) := \lim_{\rho \rightarrow 1} \widehat{h}_\epsilon^\rho(\lambda)$.

Lemma

For any $\lambda \in \mathbb{R}^d$, and for any $\delta > 0$:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} \left(\widehat{h}_\epsilon^1(\lambda) - \delta \right) \right] < \beta_{\lambda,\epsilon}(\sigma) < \exp \left[\frac{2}{\sigma^2} \left(\widehat{h}_\epsilon^1(\lambda) + \delta \right) \right] \right\} = 1.$$

Moreover,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \min \left(\text{dist} \left(x_{\beta_{\lambda,\epsilon}(\sigma)}^\sigma, B(\lambda, \epsilon) \right), \text{dist} \left(y_{\beta_{\lambda,\epsilon}(\sigma)}^\sigma, B(\lambda, \epsilon) \right) \right) \leq \delta \right\} = 1.$$

- Kramers' type laws for $c_\epsilon(\sigma) = \inf_\lambda \beta_{\lambda, \epsilon}(\sigma) = \inf\{t \geq 0 : |x_t^\sigma - y_t^\sigma| = \epsilon\}$:

Preliminary remark:

$$\widehat{h}_\epsilon^1(\lambda) = \begin{cases} \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } \min_{i=1,2} |\lambda - \lambda_i| \geq \epsilon, \\ \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } |\lambda - \lambda_1| < \epsilon, \\ \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) & \text{if } |\lambda - \lambda_2| < \epsilon. \end{cases}$$

Choosing $\epsilon < \epsilon_c$ for

$$\epsilon_c := \inf \left\{ \epsilon \in (0, \epsilon_0) : \inf_{\lambda \in B(\lambda_i, \epsilon)} \inf_{z \in \partial B(\lambda, \epsilon)} (\Psi_j(z) - \Psi_j(\lambda_j)) = \inf_{\lambda} h_\epsilon(\lambda), \quad i \neq j \in \{1, 2\} \right\},$$

then

$$\inf_{\lambda} \widehat{h}_\epsilon^1(\lambda) \approx \inf_{\lambda} h_\epsilon(\lambda),$$

$$h_\epsilon(\lambda) := \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)).$$

- Kramers' type laws for $c_\epsilon(\sigma) = \inf_\lambda \beta_{\lambda,\epsilon}(\sigma) = \inf\{t \geq 0 : |x_t^\sigma - y_t^\sigma| = \epsilon\}$:

Preliminary remark:

$$\widehat{h}_\epsilon^1(\lambda) = \begin{cases} \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } \min_{i=1,2} |\lambda - \lambda_i| \geq \epsilon, \\ \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)) & \text{if } |\lambda - \lambda_1| < \epsilon, \\ \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) & \text{if } |\lambda - \lambda_2| < \epsilon. \end{cases}$$

Choosing $\epsilon < \epsilon_c$ for

$$\epsilon_c := \inf \left\{ \epsilon \in (0, \epsilon_0) : \inf_{\lambda \in B(\lambda_i, \epsilon)} \inf_{z \in \partial B(\lambda, \epsilon)} (\Psi_j(z) - \Psi_j(\lambda_j)) = \inf_{\lambda} h_\epsilon(\lambda), \quad i \neq j \in \{1, 2\} \right\},$$

then

$$\inf_{\lambda} \widehat{h}_\epsilon^1(\lambda) \approx \inf_{\lambda} h_\epsilon(\lambda),$$

$$h_\epsilon(\lambda) := \inf_{x \in \partial B(\lambda, \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \partial B(\lambda, \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)).$$

Theorem

For any $\epsilon < \epsilon_c$ and $\delta > 0$:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{h}_\epsilon - \delta) \right] < c_\epsilon(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{h}_\epsilon + \delta) \right] \right\} = 1,$$

where

$$\underline{h}_\epsilon = \inf_{\lambda} h_\epsilon(\lambda).$$

In addition, for \mathcal{H}_ϵ the set of all minimizers λ_ϵ of h_ϵ :

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \inf_{\lambda_\epsilon \in \mathcal{H}_\epsilon} \max \left(\text{dist}(x_{c_\epsilon(\sigma)}^\sigma, B(\lambda_\epsilon, \epsilon)), \text{dist}(y_{c_\epsilon(\sigma)}^\sigma, B(\lambda_\epsilon, \epsilon)) \right) \leq \delta \right\} = 1.$$

Note: The exit-cost \underline{h}_ϵ can be achieved in more than one point, but, as $\epsilon \downarrow 0$,

$$h_\epsilon(\lambda) \rightarrow (\Psi_1(\lambda) - \Psi_1(\lambda_1)) + (\Psi_2(\lambda) - \Psi_2(\lambda_2)),$$

and \mathcal{H}_ϵ concentrates onto a single point, the minimizer of h_0 .

Corollary

For

$$h_0(\lambda) = (\Psi_1(\lambda) - \Psi_1(\lambda_1)) + (\Psi_2(\lambda) - \Psi_2(\lambda_2)),$$

$$\lambda_0 = \operatorname{argmin} h_0 = (\nabla \Psi_1 + \nabla \Psi_2)(0)^{-1},$$

and $\underline{h}_0 := h_0(\lambda_0)$, we have: for any $\delta > 0$:

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{h}_0 - \delta) \right] < c_\epsilon(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{h}_0 + \delta) \right] \right\} = 1$$

and

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \max \left(|x_{c_\epsilon(\sigma)}^\sigma - \lambda_0|, |y_{c_\epsilon(\sigma)}^\sigma - \lambda_0| \right) \leq \delta \right\} = 1.$$

Kramers' type law for the first collision time between two self-stabilizing diffusions

Under the assumptions (A) – (i) to (A) – (iv), [HIP08]: the dynamics $(X_t)_{t \geq 0}$ and $(Y_t)_{t \geq 0}$ are wellposed (in the pathwise sense) and

$$\sup_{t \geq 0} \mathbb{E}[|X_t|^p + |Y_t|^p] < \infty, \forall p \in \mathbb{Z}.$$

Moreover, [T21]: For any $\kappa > 0$, there exists a finite time T_κ and a critical diffusion σ_κ such that

$$\max_{\sigma \leq \sigma_\kappa, t \geq T_\kappa} \mathbb{E}[|X_t - \lambda_1|^2] + \mathbb{E}[|Y_t - \lambda_2|^2] \leq \kappa^2.$$

Corollary (Coupling estimate)

For any $\kappa > 0$, there exists $T_\kappa > 0$ such that

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \max_{t \geq T_\kappa} |X_t - x_{t, T_\kappa}^\sigma| + |Y_t - y_{t, T_\kappa}^\sigma| \geq \kappa \right\} = 0.$$

for

$$\begin{aligned} x_{t, T_\kappa}^\sigma &= X_{T_\kappa} + \sigma(B_t - B_{T_\kappa}) - \int_{T_\kappa}^t \{ \nabla V(x_s^\sigma) + \alpha(x_s^\sigma - \lambda_1) \} ds, \\ y_{t, T_\kappa}^\sigma &= Y_{T_\kappa} + \sigma(\tilde{B}_t - \tilde{B}_{T_\kappa}) - \int_{T_\kappa}^t \{ \nabla V(y_s^\sigma) + \alpha(y_s^\sigma - \lambda_2) \} ds, \end{aligned}$$

From the Kramers' type law for (x^σ, y^σ) to the Kramers' law for (X, Y) :

Theorem

Given $\epsilon > 0$ small enough,

$$H_\epsilon(\lambda) = \inf_{x \in \partial B(\lambda; \epsilon)} (\Psi_1(x) - \Psi_1(\lambda_1)) + \inf_{y \in \mathbb{B}(\lambda; \epsilon)} (\Psi_2(y) - \Psi_2(\lambda_2)),$$

$$\Psi_1(x) := V(x) + \frac{\alpha}{2} \|x - \lambda_1\|^2, \quad \Psi_2(y) := V(y) + \frac{\alpha}{2} \|y - \lambda_2\|^2,$$

and $\underline{H}_\epsilon = \inf H_\epsilon$, for any $\delta > 0$,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_\epsilon - \delta) \right] < C_\epsilon(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_\epsilon + \delta) \right] \right\} = 1.$$

In addition, for \mathcal{M}_ϵ the set of all minimizers λ_ϵ of $\lambda \mapsto H_\epsilon(\lambda)$, and for ϵ small enough:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \inf_{\lambda_\epsilon \in \mathcal{M}_\epsilon} \max \left(\text{dist}(X_{c_\epsilon(\sigma)}, B(\lambda_\epsilon, \epsilon)), \text{dist}(Y_{c_\epsilon(\sigma)}, B(\lambda_\epsilon, \epsilon)) \right) \leq \delta \right\} = 1.$$

Threshold: $\epsilon < \epsilon_c$ with

$$\epsilon_c = \inf \left\{ \epsilon \leq \epsilon_0 : \inf_{\lambda \in \mathbb{B}(\lambda_i; \epsilon)} \inf_{z \in \partial \mathbb{B}(\lambda; \epsilon)} \left(V(z) - V(\lambda_j) + \frac{\alpha}{2} \|z - \lambda_j\|^2 \right) = \inf_{\lambda} H_\epsilon(\lambda), \quad i \neq j \in \{1, 2\} \right\}.$$

Theorem

For any $\delta > 0$, we have

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 - \delta) \right] < C_\epsilon(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 + \delta) \right] \right\} = 1.$$

where

$$\underline{H}_0 = \min H_0(\lambda),$$

$$H_0(\lambda) = \lim_{\epsilon \rightarrow 0} H_\epsilon(\lambda) = 2V(\lambda) - V(\lambda_1) - V(\lambda_2) + \frac{\alpha}{2} |\lambda - \lambda_1|^2 + \frac{\alpha}{2} |\lambda - \lambda_2|^2.$$

Moreover, for

$$\lambda_0 := \operatorname{argmin}_\lambda H_0(\lambda),$$

it holds

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \{ |X_{C_\epsilon(\sigma)} - \lambda_0| \leq \delta \} = 1 = \lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \{ |Y_{C_\epsilon(\sigma)} - \lambda_0| \leq \delta \}.$$

Collision location:

$$\lambda_0 = \left(\nabla V + \alpha I_d \right)^{-1} (\alpha(\lambda_1 + \lambda_2)/2).$$

Example: For $V(x) = \frac{x^4}{4} - \frac{x^2}{2}$ and $\alpha > 1$,

$$\lambda_0 = 0, \quad \inf \underline{H}_0 = \alpha - 1,$$

$$\mathcal{M}_\epsilon = \left\{ \alpha - 1, 4\epsilon^4 - 2\epsilon^2 + \alpha(2\epsilon \pm 1)^2 + \frac{1}{2} \right\}.$$

Collision time for the particle systems

Particle model:

$$X_t^{i,N} = x_1 + \sigma B_t^i - \int_0^t \left\{ \nabla V(X_s^{i,N}) + \alpha \left(X_s^{i,N} - \frac{1}{N} \sum_{j=1}^N X_s^{j,N} \right) \right\} ds, \quad t \geq 0,$$

$$Y_t^{i,N} = x_2 + \sigma \tilde{B}_t^i - \int_0^t \left\{ \nabla V(Y_s^{i,N}) + \alpha \left(Y_s^{i,N} - \frac{1}{N} \sum_{j=1}^N Y_s^{j,N} \right) \right\} ds, \quad t \geq 0,$$

for $(B_t^1)_{t \geq 0}, \dots, (B_t^N)_{t \geq 0}$, and $(\tilde{B}_t^1)_{t \geq 0}, \dots, (\tilde{B}_t^N)_{t \geq 0}$, two families of independent Brownian motions.

Approximated first collision-time:

$$C_\epsilon^{i,N}(\sigma) = \inf \left\{ t \geq 0 : |X_t^{i,N} - Y_t^{i,N}| \leq \epsilon \right\}, \quad 1 \leq i \leq N.$$

Anticipated exit cost: As the potential related to each family of particles is given by

$$\Upsilon_N(\mathbf{x}^N) = \sum_{i=1}^N V(x_i) + \frac{\alpha}{2N} \sum_{i,j=1}^N |x_i - x_j|^2, \quad \mathbf{x}^N = (x_1, \dots, x_N) \in \mathbb{R}^{Nd},$$

the exit-cost related to the successive asymptotic " $\sigma \downarrow 0$ next $\epsilon \downarrow 0$ " of $C_\epsilon^{i,N}(\sigma)$:

$$\inf_{\lambda \in \mathbb{R}^{dN}} \inf_{\mathbf{x}^N \in \partial B^{i,N}(\lambda, \epsilon)} \Upsilon_N(\mathbf{x}^N) - \Upsilon_N(\lambda_1, \dots, \lambda_1) + \inf_{\lambda \in \mathbb{R}^{dN}} \inf_{\mathbf{y}^N \in \partial B^{i,N}(\lambda, \epsilon)} \Upsilon_N(\mathbf{y}^N) - \Upsilon_N(\lambda_2, \dots, \lambda_2),$$

for

$$B^{i,N}(\lambda, \epsilon) = \left\{ \mathbf{x}^N = (x_1, \cdot, x_N) \in \mathbb{R}^{dN}; x_i \in B(\lambda, \epsilon) \right\}.$$

Preliminary note

- As long as (A) hold the particle systems and their "linear" analog:

$$x_{t,T}^{i,\sigma} = X_T^{i,N} + \sigma(B_t^i - B_T^i) - \int_T^t \nabla V(x_s^{i,\sigma}) ds - \int_0^t \alpha(x_s^{i,\sigma} - \lambda_1) ds, \quad T \leq t,$$

$$y_{t,T}^{i,\sigma} = Y_T^{i,N} + \sigma(\tilde{B}_t^i - \tilde{B}_T^i) - \int_T^t \nabla V(y_s^{i,\sigma}) ds - \int_0^t \alpha(y_s^{i,\sigma} - \lambda_2) ds, \quad T \leq t.$$

are well-posed in the pathwise sense. Moreover, for all $1 \leq i \leq N$, T finite and $1 \leq p < \infty$

$$\max_{t \in [0, T]} \mathbb{E}[|X_t^{i,N}|^p + |Y_t^{i,N}|^p] < \infty.$$

- Propagation of chaos: For X^1, \dots, X^N and Y^1, \dots, Y^N , N independents copies of X_t and Y driven respectively by B^1, \dots, B^N and $\tilde{B}^1, \dots, \tilde{B}^N$,

$$\mathbb{E}[\max_{0 \leq t \leq T} |X_t^{i,N} - X_t^i|^p + \max_{0 \leq t \leq T} |Y_t^{i,N} - Y_t^i|^2] \leq \frac{C(\sigma, T)}{N}.$$

Non-uniform propagation of chaos \Rightarrow We cannot rely on the Kramers' law established in the mean-field limit situation to deal with the particle case.

\Rightarrow **Start over and apply a strategy analog to the mean-field case.**

Lemma

For any $\kappa > 0$, there exists a (non-random) time $0 \leq T_\kappa$, uniform with respect to σ , and a (non-random) finite number of particles N_κ , both finite and independent of σ , such that, for all $N \geq N_\kappa$, it holds

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \max_{t \in [T_\kappa, \exp[\frac{2}{\sigma^2}(\underline{H}_\epsilon + 2)]]} \left(\left| \frac{1}{N} \sum_{j=1}^N X_t^{j,N} - \lambda_1 \right| + \left| \frac{1}{N} \sum_{j=1}^N Y_t^{j,N} - \lambda_2 \right| \right) \leq 2\kappa \right\} = 1.$$

Proposition

For any $\xi > 0$, there exists $0 < T_\xi < \infty$ and $0 < N_\xi < \infty$ such that

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left(\sup_{T_\kappa \leq t \leq \exp[\frac{2}{\sigma^2}(\underline{H}_\epsilon + 2)]} \left\{ |X_t^{i,N} - x_{t,T_\kappa}^{i,\sigma}| + |Y_t^{i,N} - y_{t,T_\kappa}^{i,\sigma}| \right\} \geq \xi \right) = 0.$$

Kramers' law for the first collision time

(Analogous to the mean-field limit - in particular same threshold $\epsilon < \epsilon_c$).

Theorem

Let \mathcal{M}_ϵ be the set of minimizers of H_ϵ . Then, for any $\delta > 0$ and ϵ small enough, provided that N is large enough:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_\epsilon - \delta) \right] < C_\epsilon^{i,N}(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_\epsilon + \delta) \right] \right\} = 1,$$

and

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \inf_{\lambda_\epsilon \in \mathcal{M}_\epsilon} \max \left(\text{dist}(X_{C_\epsilon^{i,N}(\sigma)}^{i,N}, B(\lambda_\epsilon, \epsilon)), \text{dist}(Y_{C_\epsilon^{i,N}(\sigma)}^{i,N}, B(\lambda_\epsilon, \epsilon)) \right) \leq \delta \right\} = 1.$$

Corollary

For any $\delta > 0$, provided N is large enough:

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 - \delta) \right] < C_\epsilon^{i,N}(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 + \delta) \right] \right\} = 1.$$

Moreover, for any $1 \leq i \leq N$,

$$\lim_{\epsilon \rightarrow 0} \lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \max \left(|X_{C_\epsilon^{i,N}(\sigma)}^{i,N} - \lambda_0|, |Y_{C_\epsilon^{i,N}(\sigma)}^{i,N} - \lambda_0| \right) \leq \delta \right\} = 1.$$

Note on the one-dimensional case

In this situation, one can deal more directly with the true collision times:

$$C(\sigma) = \inf \{t \geq 0 : X_t = Y_t\}, \quad C^{i,N}(\sigma) = \inf \{t \geq 0 : X_t^{i,N} = Y_t^{i,N}\}.$$

Theorem

For any $\delta > 0$:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 - \delta) \right] < C(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 + \delta) \right] \right\} = 1.$$

Moreover, for λ_0 the minimizer of H_0 ,

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \{ |X_{C(\sigma)} - \lambda_0| \leq \delta \} = 1.$$

Theorem

For any $\delta > 0$, and N sufficiently large:

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \left\{ \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 - \delta) \right] < C^{i,N}(\sigma) < \exp \left[\frac{2}{\sigma^2} (\underline{H}_0 + \delta) \right] \right\} = 1,$$

and, for all $1 \leq i \leq N$

$$\lim_{\sigma \rightarrow 0} \mathbb{P} \{ |X_{C^{i,N}(\sigma)}^{i,N} - \lambda_0| \leq \delta \} = 1.$$

Generalizations

Our main results holds true for:

- more general self-stabilizing forces:

$$X_t = x_1 + \sigma B_t - \int_0^t \left\{ \nabla V(X_s) + \nabla F(X_s - \mathbb{E}[X_s]) \right\} ds, \quad t \geq 0,$$

$$Y_t = x_2 + \sigma \tilde{B}_t - \int_0^t \left\{ \nabla V(Y_s) + \nabla F(Y_s - \mathbb{E}[Y_s]) \right\} ds, \quad t \geq 0,$$

provided that F is a smooth function such that $F(x) = G(|x|)$ where $G : \mathbb{R} \rightarrow \mathbb{R}$ is a even polynomial function G , with a degree larger than 2, satisfying $G(0) = 0$ (i.e. framework of [T20]). Same for the related particle systems.

- Multi-wells confining potential: For instance if V admits m wells located at $\lambda_1, \dots, \lambda_m$ then, again, the Kramers' law for $C_\epsilon(\sigma)$, $C_{\epsilon,N}(\sigma)$, $C(\sigma)$ and $C_N(\sigma)$ hold and the collision λ_0 is located at

$$\left(\sum_{l=1}^m \nabla \Psi_l \right)^{-1} \left(\alpha m^{-1} \sum_{l'=1}^m \lambda_{l'} \right), \quad \Psi_l(x) = V(x) + \frac{\alpha}{2} |x - \lambda_l|^2.$$

- Random initial conditions: As long as (x_0^σ, y_0^σ) or (X_0, Y_0) are a.s. bounded (sufficient for control of moments), and the law of X_0 and Y_0 have full support on different basin of attraction of V , our main results for the self-stabilizing systems still hold true. Not so much for the particle systems.

References



P. Cattiaux, F. Delebecque, L. Pédèches.
Stochastic Cucker-Smale models: Old and new,
Annals of Applied Probability, 28 (2018), no 5, 3239-3286.



F. Cucker, S. Smale.
On the mathematics of emergence,
Japan. J. Math., 2 (2007), 197-227.



A. Dembo, O. Zeitouni.
Large deviations techniques and applications, 1998.



M. I. Freidlin and A. D. Wentzell.
Random perturbations of dynamical systems, 1998.



S. Herrmann, P. Imkeller, and D. Peithmann.
Large deviations and a Kramers' type law for self-stabilizing diffusions.
Ann. Appl. Probab., 18 (2008), no 4, 1379-1423.



J. Tugaut.
Exit-time of mean-field particles system.
ESAIM: Probability and Statistics, 24 (2020), 399-407.



J. Tugaut.
Captivity of the solution to the granular media equation.
Kinetic and Related Models, 14 (2021), no. 2, 199-209.