

# Captivity of mean-field systems\*

Julian Tugaut  
Fakultät für Mathematik  
Universität Bielefeld  
D-33615 Bielefeld  
Germany  
Email: jtugaut@math.uni-bielefeld.de

## Abstract

We investigate the exit time of some mean-field system which is related to McKean-Vlasov diffusions under convex interaction and double-well exterior force. In one hand, we show that the meta-potential which intervenes in the system admits a number of wells which tends to infinity when the number of particles tends to infinity. In the other hand, by using the convergence of the self-stabilizing processes, the phase transitions of the granular media equation, we show that there exist some traps such that the diffusion in  $\mathbb{R}^N$  can not escape when  $N$  tends to infinity. This means that in high dimension and with fixed noise, the meta-potential is not the function which drives the random dynamical system.

**Key words and phrases:** Interacting particle system ; Propagation of chaos ; Exit-time ; Non-convexity ; Free-energy ; Stationary measures.

**2000 AMS subject classifications:** primary 82C22, 60F10; secondary: 60J60, 60G10

## Introduction

We are interested in the exit times - when  $N$  tends towards infinity - of a mean-field system of the following form:

$$\left\{ \begin{array}{l} dX_t^1 = \sqrt{\epsilon} dB_t^1 - V'(X_t^1) dt - \frac{1}{N} \sum_{j=1}^N F'(X_t^1 - X_t^j) dt \\ \vdots \\ dX_t^i = \sqrt{\epsilon} dB_t^i - V'(X_t^i) dt - \frac{1}{N} \sum_{j=1}^N F'(X_t^i - X_t^j) dt \\ \vdots \\ dX_t^N = \sqrt{\epsilon} dB_t^N - V'(X_t^N) dt - \frac{1}{N} \sum_{j=1}^N F'(X_t^N - X_t^j) dt \end{array} \right. \quad (\text{I})$$

---

\*Supported by the DFG-funded CRC 701, Spectral Structures and Topological Methods in Mathematics, at the University of Bielefeld.

where the  $N$  Brownian Motions  $(B_t^i)_{t \in \mathbb{R}_+}$  are independent. We write  $X_t^i$  instead of  $X_t^{i,N,\epsilon}$  for simplifying the reading. In this paper, we will make some smoothness assumptions on the interacting potential  $F$ .

There is a huge literature about mean-field systems. Let us note some applications of this kind of system: [CDPS10] about social interactions or [CX10] about the stochastic partial differential equations. Diffusion (I) is continuous but discrete space have also been studied, particularly the Currie-Weiss model, see [BBI09] or [MP98] for example.

By introducing  $\mathcal{X}_t := (X_t^1, \dots, X_t^N)$  and  $\mathcal{B}_t := (B_t^1, \dots, B_t^N)$ , (I) corresponds to a classical diffusion in  $\mathbb{R}^N$ :

$$d\mathcal{X}_t = \sqrt{\epsilon} d\mathcal{B}_t - N \nabla \Upsilon^N(\mathcal{X}_t) dt \quad (\text{I})$$

where  $\Upsilon^N$  is called the meta-potential and is defined by:

$$\Upsilon^N(\mathcal{X}) := \frac{1}{N} \sum_{i=1}^N V(X_i) + \frac{1}{2N^2} \sum_{1 \leq i, j \leq N} F(X_i - X_j) \quad (\text{II})$$

for all  $\mathcal{X} = (X_1, \dots, X_N) \in \mathbb{R}^N$ . The motion of the process  $(\mathcal{X}_t)_{t \in \mathbb{R}_+}$  is subject to three concurrent forces. The first one is the gradient of the so-called confining potential  $V$  which acts independently on each coordinate. The second term represents the average tension of the interacting potential  $F$  between the coordinates. The third influence is some heat process  $(\sqrt{\epsilon} \mathcal{B}_t)_{t \in \mathbb{R}_+}$  which allows the particle to escape from the stable domains of the meta-potential  $\Upsilon^N$ . Let us stress that  $\epsilon$  is not necessary asymptotically small. The first two forces generate the meta-potential  $\Upsilon^N$ . We divided by  $N$  in (II) for the definition of  $\Upsilon^N$  in order to stress the influence of  $N$  in the line level. Indeed, Lemma 5.3 in [Tug10a] tells us that for all probability law  $\mu$  on  $\mathbb{R}$  absolutely continuous with respect to the Lebesgue measure, we have:

$$\Upsilon^N(X^1, \dots, X^N) \longrightarrow \int_{\mathbb{R}} \left\{ V(x) + \frac{1}{2} F * \mu(x) \right\} \mu(x) dx \quad (\text{III})$$

where  $(X^i)_{i \in \mathbb{N}^*}$  is an iid sequence of random values with law  $\mu$ . Here,  $*$  denotes the convolution.

When  $N$  tends to infinity, each particle can be seen as a diffusion in  $\mathbb{R}^N$  which satisfies the following non-linear stochastic differential equation:

$$\begin{cases} X_t = X_0 + \sqrt{\epsilon} B_t - \int_0^t V'(X_s) ds - \int_0^t F' * u_s(X_s) ds \\ u_s = \mathcal{L}(X_s) \end{cases} \quad (\text{IV})$$

The own law of this process intervenes in the equation. Consequently, it is non-markovian. We call it a self-stabilizing process. Let us give briefly some of the previous works on these diffusions (IV). For the existence problem, see [McK67, BRTV98, CGM08, Mel96, HIP08, Tug10a]. Also, in [McK67], the

author proved that the law of the solution  $du_t$  admits a  $C^\infty$ -continuous density  $u_t$  with respect to the Lebesgue measure since  $t > 0$  and the density satisfies the following non-linear parabolic partial differential equation:

$$\frac{\partial}{\partial t} u_t(x) = \frac{\partial}{\partial x} \left\{ \frac{\epsilon}{2} \frac{\partial}{\partial x} u_t(x) + u_t(x) \left( V'(x) + F' * u_t(x) \right) \right\}. \quad (\text{V})$$

This permits to study the stationary measure(s) of the process (IV). Particularly, in [HT10a], it has been proved that there is non-uniqueness of the stationary measures for  $\epsilon$  small enough. The exact number of stationary measures for  $\epsilon$  non necessary small has been the subject of [HT10b, HT09, Tug11]. By knowing the exact number of stationary measures, we can study the convergence in long-time behavior. In the convex case, see [BRV98, Mal03, Ver06, BCCP98, CMV03, CGM08].

The long-time behavior of the law  $u_t$  in the more difficult non-convex case has been proved in [Tug10b]:  $u_t$  converges weakly towards a stationary measure under assumptions easy to verify. For doing this, the main tool is the following free-energy:

$$\Upsilon_\epsilon(u) := \int_{\mathbb{R}} \left\{ \frac{\epsilon}{2} \log(u(x)) + V(x) + \frac{1}{2} F * u(x) \right\} u(x) dx \quad (\text{VI})$$

for all the measures  $du$  which are absolutely continuous with respect to the Lebesgue measure. We can note that  $du_t$  - defined in (IV) - satisfies this hypothesis since  $t > 0$ , see [HT10a]. We can observe that the second term of  $\Upsilon_\epsilon$  is linked to the meta-potential  $\Upsilon^N$  by the limit (III).

The link between the self-stabilizing processes and the mean-field system for  $N$  tending to  $+\infty$  is called the propagation of chaos, see [Szn91] under Lipschitz properties ; [BRTV98] if  $V$  is a constant ; [Mal01, Mal03] when both potentials are convex ; [BAZ99] for a more precise result ; [BGV07], [DPdH96] or [DG87] for a sharp estimate ; [CGM08] for a uniform result in time in the non-uniformly convex case. Also, a half-uniform propagation of chaos has been proved in [Tug11] when we restrict to a symmetric initial measure and when  $V$  is a double-well potential.

Let us note also some works about the propagation of chaos with different hypotheses about the dynamic or the phase space: [Gra90, Gra92, Der03, JM08].

Exit-time of diffusions have already been studied when the dimension is fixed and when the coefficient diffusion  $\sqrt{\epsilon}$  tends to 0. Indeed, for classical diffusions, Freidlin and Wentzell (see for example [DZ10, FW98]) proved a Kramer's type law theorem. In the case of the mean-field system (I), the exit time of an open set  $\mathcal{O} \subset \mathbb{R}^N$  which contains at least one well of the meta-potential  $\Upsilon^N$  is exponentially equivalent to  $\exp \left[ \frac{2N}{\epsilon} H \right]$  with  $H := \sup_{x,y \in \mathcal{O}} (\Upsilon^N(x) - \Upsilon^N(y))$ . However, in this paper, we do not want  $\epsilon$  to be small but  $N$  to be big and we can not apply this method even in the small-noise asymptotic since a commutation between the two limits is not possible.

The global method that we will use for studying the exit-time consist on using

the propagation of chaos, the convergence established in [Tug10b] and the result about the phase transition proved in [Tug11]. Indeed, let us recall that it has been proved in [DG87] that the empirical law of the mean-field system satisfies a large deviations principle with a rate function which depends on the law of Diffusion (IV). Consequently, the long-time behavior of  $\mathcal{L}(X_t)$  provides some consequences on the exit time for the particle system (I).

The paper is organized as follows. First, the assumptions and the notations are presented in first section with some of the results about self-stabilizing diffusions and two simple lemmas which will be used subsequently. The second section deals with the potential geometry, particularly the number of wells. Propagation of chaos is proved in third section then used for obtaining the main results, in particular the fact that even when there are  $2^N(1 - o(1))$  wells, there are at most  $o(2^N)$  steady states. For concluding the introduction, we write the statements of the three main results. Set  $V(x) := \frac{x^4}{4} - \frac{x^2}{2}$  and  $F(x) := \frac{\alpha}{2}x^2$  with  $\alpha > 0$ . We have:

**Big number of wells** *If  $\alpha < \frac{2}{3}$ . Then  $\Upsilon^N$  admits  $2^N(1 - o(1))$  wells.*

**Small number of steady states** *Set  $\mathcal{X} \in \mathbb{R}^N$  and  $\mathcal{X}_t$  the diffusion (I) starting with  $\mathcal{X}$ . Set  $\kappa > 0$ . There exists  $\epsilon_0 > 0$  such that for all  $\epsilon \in ]0; \epsilon_0[$ , there exists  $T_\kappa \geq 0$  such that for all  $t > 0$ , we have:*

$$\lim_{N \rightarrow +\infty} \mathbb{P} \left[ \min \left\{ p(s); 1 - p(s); \left| \frac{1}{2} - p(s) \right| \right\} \leq \kappa, \forall T \leq s \leq T + t \right] = 1$$

where  $p(s) := \frac{1}{N} \{i \in \llbracket 1; N \rrbracket \mid X_s^i > 0\}$ .

**Stability of the positive half-space** *Set  $\epsilon$  small enough and a law  $\mu_0$  such that  $\int_{\mathbb{R}} V(x)\mu_0(x)dx + \frac{\alpha}{2}\text{Var}(\mu_0) < -\frac{\max\{1-\alpha; 0\}^2}{4}$  and  $\mathbb{E}(\mu_0) > 0$ . We consider a sequence of iid random values with law  $\mu_0: (X_0^i)_{i \in \mathbb{N}^*}$  and  $\mathcal{X}_t$  the diffusion (I) starting with  $\mathcal{X} := (X_0^1, \dots, X_0^N)$ . Then, for all  $t \geq 0$ , we have:*

$$\lim_{N \rightarrow +\infty} \mathbb{P} \left\{ \sum_{i=1}^N X_t^i > 0 \mid \forall 0 \leq s \leq t \right\} = 1.$$

## 1 Preliminaries

In this paper, we take  $V(x) := \frac{x^4}{4} - \frac{x^2}{2}$  and  $F(x) := \alpha \frac{x^2}{2}$  with  $\alpha \in \mathbb{R}_+^*$ . Let us recall that the value  $\sup_{z \in \mathbb{R}} -V''(z)$  plays an important role in [HT10a, HT10b, HT09]. Here, we know directly that  $\sup_{x \in \mathbb{R}} -V''(x) = 1$ . We could consider more general assumptions, the same than the ones in [Tug10b] and [Tug11]. But, all the mathematical difficulties of the present problem are present in this case. Considering these two functions instead of general polynomial ones will permit to avoid some technical and tedious computations, especially in the next section. Indeed, the main results (see Section 4) only need a result of propagation of chaos (see Section 3) and a result of convergence in long-time

(see [Tug10b]).

Furthermore, by adapting Theorem 2.1 in [Tug11] to our problem, we have the following result:

**Proposition 1.1.** *For all  $\alpha > 0$ , there exists  $\epsilon_c(\alpha) \in \mathbb{R}_+$  such that:*

- *For all  $\epsilon \geq \epsilon_c(\alpha)$ , Diffusion (IV) admits a unique stationary measure:  $u_\epsilon^0$ .*
- *For all  $\epsilon < \epsilon_c(\alpha)$ , Diffusion (IV) admits exactly three stationary measures:  $u_\epsilon^0$ ,  $u_\epsilon^+$  and  $u_\epsilon^-$  with  $\pm \int_{\mathbb{R}} x u_\epsilon^\pm(x) dx > 0$ .*

Moreover, the critical value  $\epsilon_c(\alpha)$  is the unique solution of the equation:

$$\int_{\mathbb{R}_+} \left( x^2 - \frac{1}{2\alpha} \right) \exp \left[ (1 - \alpha) x^2 - \frac{\epsilon}{2} x^4 \right] dx = 0. \quad (1.1)$$

We have then a simple representation of the threshold between the uniqueness and the thirddness of the stationary measure(s) of Diffusion (IV):

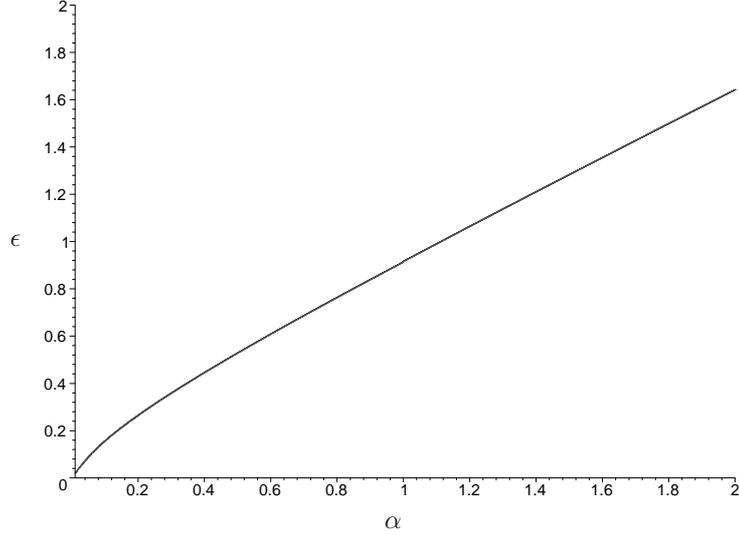


Figure 1: Critical value  $\epsilon_c(\alpha)$

All the measure  $u_0$  which will be considered in the following satisfy the three following hypotheses:

**(AC)**  $u_0$  is absolutely continuous with respect to the Lebesgue measure. We call  $u_0$  the density for simplicity.

**(FM)**  $\int_{\mathbb{R}} x^{32} u_0(x) dx < \infty$ .

**(FE)**  $\int_{\mathbb{R}} \log(u_0(x)) u_0(x) dx < \infty$ .

The assumptions (AC) and (FE) guarantee that  $\Upsilon_\epsilon(u_0) < \infty$ . The condition (FM) simply expresses the existence of a strong solution on  $\mathbb{R}_+$  for Equation (IV). Indeed, by Theorem 2.12 in [HIP08], we know that there is a solution if  $\int_{\mathbb{R}} x^{8q^2} u_0(x) dx < \infty$  where  $q := \max \left\{ \frac{\deg(V)}{2}; \frac{\deg(F)}{2} \right\}$  if  $V$  and  $F$  are polynomial functions.

The three assumptions imply the existence of  $M_0 > 0$  such that

$$\max_{j \in \llbracket 1; 32 \rrbracket} \sup_{t \in \mathbb{R}_+} \mathbb{E} \left[ |X_t|^j \right] \leq M_0. \quad (1.2)$$

We deduce immediately that the family  $(u_t)_{t \in \mathbb{R}_+}$  is tight. Also, we have the convergence of  $u_t$  towards a stationary measure  $u_\epsilon$ , see [Tug10b].

Let us recall the main results of Section 2 in [HT10b] (Theorem 2.1 and Theorem 2.4), about the small-noise asymptotic of the stationary measures:

**Proposition 1.2.** *The law  $u_\epsilon^+$  converges weakly towards  $\delta_1$  and the law  $u_\epsilon^-$  converges weakly towards  $\delta_{-1}$ .*

*The law  $u_\epsilon^0$  converges weakly towards  $\frac{1}{2}\delta_{x_\alpha} + \frac{1}{2}\delta_{-x_\alpha}$  with*

$$x_\alpha := \begin{cases} \sqrt{1-\alpha} & \text{if } \alpha \leq 1 \\ 0 & \text{if } \alpha \geq 1 \end{cases}.$$

We present first some general notations. Let us note that for each  $x \in \mathbb{R}$ ,  $\mathcal{E}(x)$  denotes here the unique integer such that  $x - 1 < \mathcal{E}(x) \leq x$  and  $C_n^p$  is the binomial coefficient  $C_n^p := \frac{n!}{p!(n-p)!}$ .

For all the measures  $\mu_1$  and  $\mu_2$ ,  $\mathbb{W}_2(\mu_1, \mu_2)$  is the Wasserstein distance between  $\mu_1$  and  $\mu_2$ :

$$\mathbb{W}_2(\mu_1, \mu_2) := \sqrt{\inf_{\{\mathcal{L}(X)=\mu_1, \mathcal{L}(Y)=\mu_2\}} \mathbb{E} \left\{ |X - Y|^2 \right\}}.$$

Now, we present some notations linked to the spaces  $\mathbb{R}^N$ . By convention, we consider that  $\mathbb{R}^\infty$  is the set of the measures on  $\mathbb{R}$ . In the following,  $\mathcal{X}$  is an arbitrary element of  $\mathbb{R}^N$  and the  $i$ -th coordinate is written  $X_i$ , when  $N < \infty$ . Let us introduce some definitions:

**Definition 1.3. 1)** *For all  $\mathcal{X} := (X_1, \dots, X_N) \in \mathbb{R}^N$ , we consider the norm  $\|\mathcal{X}\|_N := \sqrt{\frac{\sum_{i=1}^N X_i^2}{N}}$ .*

**2)** *For all  $r > 0$ , we introduce  $\mathbb{B}_r^N := \left\{ \mathcal{X} \in \mathbb{R}^N \mid \|\mathcal{X}\|_N \leq r \right\}$ .*

**3)** *We consider also  $\mathbb{E}_+^N := \left\{ \mathcal{X} \in \mathbb{R}^N \mid \sum_{i=1}^N X_i > 0 \right\}$ .*

**4)** *For each  $\mathcal{X}_0 \in \mathbb{R}^N$ , we call  $\mathcal{X}_t$  the mean-field system (I) starting by  $\mathcal{X}_0$  and  $X_t^i$  the  $i$ -th coordinate of  $\mathcal{X}_t$ .*

We introduce now the notion of signature.

**Definition 1.4.** Set  $N \in \mathbb{N}$  and  $\mathcal{X} \in \mathbb{R}^N$ . We say that  $\mathcal{X}$  has the signature  $(p, 1-p)$  with  $p \in \frac{1}{N} \llbracket 1; N \rrbracket$  if  $\#\{i \in \llbracket 1; N \rrbracket \mid X_i > 0\} = pN$ .

Set a sequence  $\mathcal{X} \in \mathbb{R}^{\mathbb{N}}$ . We say that  $\mathcal{X}$  has the signature  $(p, 1-p)$  with  $p \in [0; 1]$  if  $\lim_{N \rightarrow +\infty} \frac{\#\{i \in \llbracket 1; N \rrbracket \mid X_i > 0\}}{N} = p$ . Obviously, if  $\mu \in \mathbb{R}^{\infty}$  is a measure, we say that it has the signature  $(p, 1-p)$  if  $\mu(\{0\}) = 0$  and  $\mu(\mathbb{R}_+) = p$ .

Immediately, if  $\mathcal{X}$  is a sequence of iid random values with law  $\mu_0$  absolutely continuous with respect to the Lebesgue measure,  $\mathcal{X}$  has the signature  $(\int_{\mathbb{R}_+} \mu_0(x) dx, \int_{\mathbb{R}_-} \mu_0(x) dx)$ .

**Definition 1.5.** Set  $p \in [0; 1]$  and  $N \in \mathbb{N} \cup \{+\infty\}$ . We call  $\mathbb{S}_p^N$  the set of the elements  $\mathcal{X} \in \mathbb{R}^N$  which have the signature  $(p, 1-p)$ .

**Remark 1.6.** The particularity of the random dynamical system that we consider is its invariance for each element  $\sigma \in \mathcal{S}_N$  where  $\mathcal{S}_N$  is the set of all the permutations of the set  $\llbracket 1; N \rrbracket$ . Consequently, we will not work on  $\mathbb{R}^N$  but on  $\mathbb{R}^N / \mathcal{S}_N$ . For simplifying the reading, we will not specify that we consider class of equivalence instead of elements of  $\mathbb{R}^N$ . Particularly, for  $\mathcal{X}, \mathcal{Y} \in \mathbb{R}^N$ , the expression  $\sum_{i=1}^N (X_i - Y_i)^2$  will be in fact a notation which corresponds to  $\inf_{\sigma \in \mathcal{S}_N} \sum_{i=1}^N (X_{\sigma(i)} - Y_i)^2$ .

Let us stress that this last expression corresponds to the Wasserstein distance between two measures with support containing  $N$  elements at most. Also, we can remark that  $\mathbb{E}_+^N$ ,  $\mathbb{B}_r^N$  and  $\mathbb{S}_p^N$  are invariant under the actions of any permutations so we can use these three sets without paying attention on the difference between  $\mathbb{R}^N$  and  $\mathbb{R}^N / \mathcal{S}_N$ .

Since we use the classes of equivalence of  $\mathbb{R}^N$  with respect to the permutations instead of  $\mathbb{R}^N$  itself, we introduce some particular classes.

**Definition 1.7. 1)** Set  $x \in \mathbb{R}$ , we define the vector  $\bar{x} \in \mathbb{R}^N$  as a pure state with coordinates equal to  $x$ :  $\bar{x} := (x, \dots, x)$ .

**2)** Set  $a, b \in \mathbb{R}$  and  $p \in ]0; 1[$ . The class of vectors  $(a, b, p)$  denotes all the bi-state vectors with  $\mathcal{E}(pN)$  coordinates equal to  $a$  and  $N - \mathcal{E}(pN) \neq \mathcal{E}((1-p)N)$  coordinates equal to  $b$ . Let us note that this class contains exactly  $C_N^{\mathcal{E}(pN)}$  elements of  $\mathbb{R}^N$ .

In order to conclude the preliminaries, we put two lemmas which will be used in the next section. The first one is about linear algebra:

**Lemma 1.8.** Let  $a, b \in \mathbb{R}$ ,  $c \in \mathbb{R}_-$ ,  $N \geq 1$  and  $k \in \llbracket 1; N-1 \rrbracket$ . Set  $I_k$  the identity matrix with size  $k$  and  $J_k$  the matrix whose each coordinate is equal to 1 and with size  $k$ . We define in the same way  $I_{N-k}$  and  $J_{N-k}$ . Finally, we define the following matrix per blocks:

$$M := \begin{pmatrix} (a-c)I_k + cJ_k & (c) \\ (c) & (b-c)I_{N-k} + cJ_{N-k} \end{pmatrix}.$$

If  $a + (N-1)c > 0$  and  $b + (N-1)c > 0$  then  $M > 0$ .

*Proof.* Let us assume  $a + (N - 1)c > 0$  and  $b + (N - 1)c > 0$ . By definition of  $M$ , for all  $\mathcal{X} \in \mathbb{R}^N$  with  $\mathcal{X} \neq \bar{0}$ , we have:

$$\begin{aligned} \langle \mathcal{X}; M\mathcal{X} \rangle &= \sum_{i=1}^k (a + (N - 1)c)X_i^2 + \sum_{i=k+1}^N (b + (N - 1)c)X_i^2 \\ &\quad - N^2c \left\{ \frac{1}{N} \sum_{i=1}^N X_i^2 - \left( \frac{1}{N} \sum_{i=1}^N X_i \right)^2 \right\}. \end{aligned}$$

It achieves the proof since  $c \leq 0$ .  $\square$

The second result is combinatorial and is just the direct application of Sterling formula:

**Lemma 1.9.** *Set  $\delta \in ]0; \frac{1}{2}[$ . Then  $\lim_{N \rightarrow +\infty} 2^{-N} \sum_{n=(\frac{1}{2}-\delta)N}^{n=(\frac{1}{2}+\delta)N} C_N^n = 1$ .*

## 2 Potential geometry

This section is devoted to the geometry of the meta-potential  $\Upsilon^N$ . We note that it is immediate that  $\bar{-1}$ ,  $\bar{1}$  and  $\bar{0}$  are critical points of  $\Upsilon^N$ . First, we will study the geometry when  $\alpha \geq 1$  that is to say in the synchronized case.

**Proposition 2.1.** *Set  $N \geq 2$  and  $\alpha \geq 1$ . Then, the meta-potential  $\Upsilon^N$  admits exactly three critical points:  $\bar{-1}$ ,  $\bar{1}$  and  $\bar{0}$ . The first two ones are wells and if  $\alpha > 1$ , the point  $\bar{0}$  is a saddle whose the signature of the Hessian is  $(N - 1, 1)$ .*

*Proof. Step 1.* Let us prove that there are exactly three critical points if  $\alpha \geq 1$ . For all  $1 \leq i \leq N$ , the derivative of the meta-potential with respect to  $x_i$  is

$$\frac{\partial}{\partial x_i} \Upsilon^N(x_1, \dots, x_N) = \frac{1}{N} \left\{ V'(x_i) + \alpha x_i - \frac{\alpha}{N} \sum_{j=1}^N x_j \right\}.$$

Set a critical point  $\mathcal{X}$ . We deduce  $\rho(x_i) = \rho(x_j)$  for all the indexes  $i$  and  $j$  with  $\rho(x) := V'(x) + \alpha x$ . By recalling  $\rho'(x) = V''(x) + \alpha \geq 0$  for all  $x \in \mathbb{R}$  according to the hypothesis  $\alpha \geq 1$ . We deduce directly  $x_i = x_j$  for all the indexes  $i$  and  $j$ . Consequently, there exists  $x \in \mathbb{R}$  such that  $\mathcal{X} = \bar{x}$ .

However, since the derivative with respect to  $x_1$  is equal to 0, we obtain  $V'(x) + \alpha x - \frac{\alpha}{N} \sum_{j=1}^N x = 0$  which implies  $V'(x) = 0$ . Then  $x \in \{-1, 0, 1\}$ .

**Step 2.** We compute the Hessian of  $\Upsilon^N$  on the points  $\bar{x}$ :

$$\begin{aligned} \frac{\partial^2}{\partial x_i^2} \Upsilon^N(x, \dots, x) &= \frac{1}{N} \left\{ V''(x) + \alpha \left( 1 - \frac{1}{N} \right) \right\} \\ \text{and } \frac{\partial^2}{\partial x_i \partial x_j} \Upsilon^N(x, \dots, x) &= -\frac{\alpha}{N^2}. \end{aligned}$$

We apply Lemma 1.8 and we deduce that the Hessian is strictly positive if it is in  $\overline{\pm 1}$ . And, a simple algebra calculus tells us that the two eigenvalues in  $\overline{0}$  are  $\alpha - 1 > 0$  associated to an eigenspace of dimension  $N - 1$  and  $-1$  which achieves the proof.  $\square$

**Remark 2.2.** We can remark that  $\overline{1}$  and  $\overline{-1}$  are wells of the meta-potential  $\Upsilon^N$  even if  $\alpha < 1$ .

According to Proposition 1.3 in [Tug11], the diffusion (IV) admits an outlying stationary measure around 0 for  $\epsilon$  small enough if we have

$$V(x) + \frac{\alpha}{2}x^2 > 0 \quad \text{for all } x \neq 0. \quad (2.1)$$

If  $\alpha > 1$ , (2.1) holds which proves the existence of an outlying stationary measure near  $\delta_0$  for  $\epsilon$  small enough. But,  $\overline{0}$  is *never* a wells of  $\Upsilon^N$  when  $\alpha > 1$ .

*This points out the importance of the entropy and  $\epsilon$  since there is no correspondance between the wells of  $\Upsilon^N$  and the stationary measures of (IV).*

Now we will study the critical points when  $\alpha \leq 1$ .

**Theorem 2.3.** *If  $\frac{2}{3} \leq \alpha < 1$ , the meta-potential  $\Upsilon^N$  admits exactly  $2^N$  critical points but only two wells:  $\overline{1}$  and  $\overline{-1}$ .*

*If  $\alpha < \frac{2}{3}$ , the meta-potential  $\Upsilon^N$  admits exactly  $2^N$  critical points and  $2^N(1 - o(1))$  of these critical points are wells of  $\Upsilon^N$ .*

Let us remark that the bifurcation  $\alpha = \frac{2}{3}$  already appeared in the proof of Theorem 3.2 in [HT10a] since  $\frac{2}{3} = -\frac{V'(c)}{c}$  where  $c$  is the unique positive real such that  $V''(c) = 0$ .

*Proof.* In order to stress the importance of  $\alpha$ , we will write  $\Upsilon_\alpha^N$  instead of  $\Upsilon^N$  in this proof.

**Step 1.** First, we study the number of critical points. The partial derivative with respect to the coordinate  $x_i$  is:

$$\frac{\partial}{\partial x_i} \Upsilon_\alpha^N(x_1, \dots, x_N) = \frac{1}{N}x_i^3 - \frac{1}{N}(1 - \alpha)x_i - \alpha \frac{1}{N^2} \sum_{j=1}^N x_j.$$

Since the third derivative of the application  $x \mapsto x^3 - (1 - \alpha)x$  is nonnegative, we deduce that for all  $\mathcal{X} \in \mathbb{R}^N$  such that  $\nabla \Upsilon_\alpha^N(\mathcal{X}) = \overline{0}$ ,  $\mathcal{X}$  has the form  $\overline{x}$  or  $(a_1, a_2, p)$ . Let  $p \in \frac{1}{N}[[1; N]]$ . We have now to solve

$$a_1^3 - a_2^3 + (1 - \alpha)(a_2 - a_1) = 0 \quad (2.2)$$

$$a_1^3 - (1 - \alpha)a_1 = \alpha(pa_1 + (1 - p)a_2). \quad (2.3)$$

In the case where  $a_1 = a_2 = a$ , we find directly  $a \in \{1, 0, -1\}$  and  $p$  does not have any importance. If  $a_2 \neq a_1$ , we can divide (2.2) by  $a_2 - a_1$  before solving

it. Then we put  $p := \frac{a_2 - a_2^3}{\alpha(a_2 - a_1)}$  which shall be in  $\frac{1}{N} \llbracket 1; N \rrbracket$ . Dividing (2.2) by  $a_2 - a_1$ , we obtain:

$$a_2^2 + a_1 a_2 + a_1^2 = 1 - \alpha \quad (2.4)$$

which implies the existence of  $t \in ]-\pi; \pi]$  such that

$$a_2(t) = 2\sqrt{\frac{1-\alpha}{3}} \cos\left(t - \frac{\pi}{6}\right) \quad \text{and} \quad a_1(t) = -2\sqrt{\frac{1-\alpha}{3}} \cos\left(t + \frac{\pi}{6}\right).$$

The equations (2.4) and (2.3) imply that  $a_1 a_2 < 0$ . Indeed, let us assume that  $a_1 \geq 0$  and  $a_2 \geq 0$ . Since  $p \geq 0$ , (2.3) provides  $a_1 > \sqrt{1-\alpha}$  and the same holds with  $a_2$ . We obtain then  $a_1^2 + a_1 a_2 + a_2^2 \geq 2(1-\alpha)$  which is impossible according to (2.4). Without loss of generality, we assume  $a_1(t) < 0$  and  $a_2(t) > 0$ . We deduce that  $t \in I := ]-\frac{\pi}{3}; \frac{\pi}{3}[$ . After computation, we find:

$$p(t) = \frac{1}{2} + \frac{\tan(t)}{2\alpha\sqrt{3}} \left\{ (3\alpha - 2) + \frac{8(1-\alpha)}{3} \sin^2(t) \right\}.$$

We remark that  $p(-\frac{\pi}{3}) = 0$  and  $p(\frac{\pi}{3}) = 1$ . Also, a simple study proves that the function  $t \mapsto p(t)$  is not in  $[0; 1]$  if  $t \in ]-\frac{\pi}{2}; -\frac{\pi}{3}[ \cup ]\frac{\pi}{3}; \frac{\pi}{2}[$ .

Then, for all  $N \in \mathbb{N}^*$ , for all  $k \in \llbracket 0; N \rrbracket$ , there exists  $t_k \in I$  such that  $(a_1(t_k); a_2(t_k); \frac{k}{N})$  is a critical point of the meta-potential  $\Upsilon_\alpha^N$ . Since  $\bar{1}$  and  $-\bar{1}$  are also critical points of  $\Upsilon^N$ , we deduce that the meta-potential admits exactly  $2^N$  critical points.

**Step 2.** Now, we study the Hessian in these critical points. We compute the second derivatives of  $\Upsilon_\alpha^N$  in a point  $(a_1(t_k); a_2(t_k); \frac{k}{N})$ :

$$\begin{aligned} \frac{\partial^2}{\partial x_i^2} \Upsilon_\alpha^N \left( a_1(t_k); a_2(t_k); \frac{k}{N} \right) &= \frac{V''(a_1(k))}{N} + \frac{\alpha}{N} - \frac{\alpha}{N^2} \quad \forall i \in \llbracket 1; k \rrbracket, \\ \frac{\partial^2}{\partial x_i^2} \Upsilon_\alpha^N \left( a_1(t_k); a_2(t_k); \frac{k}{N} \right) &= \frac{V''(a_2(k))}{N} + \frac{\alpha}{N} - \frac{\alpha}{N^2} \quad \forall i \in \llbracket k+1; N \rrbracket \\ \frac{\partial^2}{\partial x_i \partial x_j} \Upsilon_\alpha^N \left( a_1(t_k); a_2(t_k); \frac{k}{N} \right) &= -\frac{\alpha}{N^2} \quad \forall i, j \in \llbracket 1; N \rrbracket \quad i \neq j. \end{aligned}$$

By applying Lemma 1.8, we deduce that  $V''(a_1(t_k)) > 0$  and  $V''(a_2(t_k)) > 0$  implies the strict convexity of  $\Upsilon_\alpha^N$  in  $(a_1(t_k); a_2(t_k); \frac{k}{N})$ . It remains then to study the values  $t$  such that  $V''(a_1(t)) > 0$  and  $V''(a_2(t)) > 0$ . This is equivalent to

$$\cos\left(t - \frac{\pi}{6}\right) > \frac{1}{2\sqrt{1-\alpha}} \quad \text{and} \quad \cos\left(t + \frac{\pi}{6}\right) > \frac{1}{2\sqrt{1-\alpha}}.$$

These two inequalities hold if and only if  $t \in I_\alpha := \bigcap ]\frac{\pi}{6} - t(\alpha); t(\alpha) - \frac{\pi}{6}[$  with  $t(\alpha) := \arccos\left(\frac{1}{2\sqrt{1-\alpha}}\right)$ . And,  $\frac{\pi}{6} < t(\alpha)$  is equivalent to  $\alpha < \frac{2}{3}$ . Let us remark that  $\alpha > \frac{3}{4}$  would imply  $1 < \frac{1}{2\sqrt{1-\alpha}}$  and then we would not be able to define

$t(\alpha)$ . The function  $p$  in this interval is decreasing when  $\alpha < \frac{2}{3}$ . Indeed, a computation provides:

$$p'(t) = -\frac{16(1-\alpha)\sin^4(t) - 24(1-\alpha)\sin^2(t) + 3(2-3\alpha)}{6\alpha\sqrt{3}\cos^2(t)}.$$

So  $p'(t) \geq 0$  if and only if

$$\frac{3\sqrt{1-\alpha} - \sqrt{3}}{4\sqrt{1-\alpha}} \leq \sin^2(t) \leq \frac{3\sqrt{1-\alpha} + \sqrt{3}}{4\sqrt{1-\alpha}}.$$

But, for  $t \in ]\frac{\pi}{6} - t(\alpha); t(\alpha) - \frac{\pi}{6}[$ , we have

$$\sin^2(t) \leq \sin^2\left(t(\alpha) - \frac{\pi}{6}\right) = \frac{5 - 6\alpha - \sqrt{9 - 12\alpha}}{8(1-\alpha)}.$$

However, the inequality  $\frac{5-6\alpha-\sqrt{9-12\alpha}}{8(1-\alpha)} \geq \frac{3\sqrt{1-\alpha}-\sqrt{3}}{4\sqrt{1-\alpha}}$  is equivalent to  $\alpha \geq \frac{2}{3}$ . Then, the function  $p$  is a bijection from  $I_\alpha$  to  $]p(\alpha); \overline{p(\alpha)}[ \subset [0; 1]$ . We can observe that  $]p(\alpha); \overline{p(\alpha)}[ \ni p(0) = \frac{1}{2}$ . By applying Lemma 1.9, we deduce that the number of wells is equivalent to  $2^N$  when  $N$  tends to infinity.  $\square$

We recover then a result previously obtained in [BFG07]. Even if there are  $2^N(1-o(1))$  wells when  $\alpha < \frac{2}{3}$  and  $2^N(1-o(1))$  critical points even if  $\frac{2}{3} \leq \alpha < 1$ , we will state in the following that the number of classes of steady states for the dynamic in the mean-field system (I) does not depend on  $N$ .

### 3 Exit times

We will begin to state a weak result of propagation of chaos. In other words, we will prove on a finite interval of time  $[0; T]$ , when  $N$  tends towards infinity that the behavior of each particle of (I) is like the one of a self-stabilizing process (IV). We recall their definition:

$$X_t^i = X_0^i + \sqrt{\epsilon}B_t^i - \int_0^t V'(X_s^i)ds - \int_0^t \frac{1}{N} \sum_{j=1}^N F'(X_s^i - X_s^j)ds \quad (\text{I})$$

$$\text{and } \overline{X}_t^i = X_0^i + \sqrt{\epsilon}B_t^i - \int_0^t V'(\overline{X}_s^i)ds - \int_0^t F' * u_s^i(\overline{X}_s^i)ds, \quad (\text{IV})$$

where  $B_t^1, \dots, B_t^N$  are  $N$  independent Brownian Motions.

**Proposition 3.1.** *Set a probability measure  $\mu_0$  which satisfies the hypothesis (FM). Set  $X_0^1, \dots, X_0^N$   $N$  iid random values with law  $\mu_0$ . Set  $T > 0$ . Then, there exists  $C, K > 0$  such that:*

$$\max_{1 \leq i \leq N} \mathbb{E} \left\{ \sup_{t \in [0; T]} |X_t^i - \overline{X}_t^i|^2 \right\} \leq \frac{C}{N} \exp[K T]. \quad (3.1)$$

*Proof.* We proceed like in Step 3 in the proof of Proposition 2.2 in [Tug11] but we consider  $V$  and  $F$  instead of the small modifications of Step 2. Here  $V(x) = \frac{x^4}{4} - \frac{x^2}{2}$  then Step 3.3 is modified and consequently, we obtain

$$\frac{d}{dt}\zeta_i(t) \leq 2\zeta_i(t) + \frac{2C}{\sqrt{N}}\sqrt{\zeta_i(t)}$$

where  $\zeta_i(t) := \mathbb{E} \left[ \left| X_t^i - \overline{X}_t^i \right|^2 \right]$  and  $C$  depends only of the first 32 moments of  $\mu_0$ . We obtain immediately  $\frac{d}{dt}\zeta_i(t) \leq (C+2)\zeta_i(t) + \frac{C}{N}$ . As  $\zeta_i(0) = 0$ , we deduce after applying the Grönwall lemma:

$$\zeta_i(t) \leq \frac{C}{N} \exp [(C+2)t] .$$

Taking the supremum on the interval  $[0; T]$  achieves the proof. We can put the supremum in the expectation by conditioning with the events  $\{\mathcal{T}(\omega) = t\}$  where  $\mathcal{T}$  is defined by the relation  $\left| X_{\mathcal{T}}^i - \overline{X}_{\mathcal{T}}^i \right| = \sup_{t \in [0; T]} \left| X_t^i - \overline{X}_t^i \right|$ .  $\square$

This result is not uniform with respect to the time. Indeed, the factor  $\exp [(C+2)T]$  tends to infinity when  $T$  tends to infinity. We will now prove that there is no uniform with respect to the time propagation of chaos.

**Definition 3.2.** *There is propagation of chaos of the system (I) to the system (IV) if there exists a positive function  $\eta$  which vanishes when  $N$  tends towards  $+\infty$  such that*

$$\sup_{t \geq 0} \mathbb{E} \left\{ \left| X_t^i - \overline{X}_t^i \right|^2 \right\} \leq \eta(N) \quad \text{for all } 1 \leq i \leq N . \quad (3.2)$$

This uniform propagation of chaos has been stated and used in the convex case in [CGM08] for obtaining a convergence of the process. However, in this non-convex case, we will see that there is no uniformity possible.

**Proposition 3.3.** *Let us assume a uniform in time propagation of chaos then the diffusion (IV) admits at most one stationary measure.*

*Proof.* The proof is similar to Step 2. and Step 4. in the proof of Proposition 2.2 in [Tug11] with  $V$  and  $F$  instead of  $V_0$  and  $F_0$ . Indeed, in the two majorations (2.10) and (2.11) in [Tug11], we did not exploit the exact higher-bound but only the fact that we have the uniform propagation of chaos with respect to the time.  $\square$

**Remark 3.4.** *We proved in Theorem 2.1 in [Tug11] that for  $\epsilon$  which satisfies  $\int_{\mathbb{R}_+} (x^2 - \frac{1}{2\alpha}) \exp [(1-\alpha)x^2 - \frac{\epsilon}{2}x^4] dx > 0$ , the diffusion (IV) admits exactly three stationary measures. Consequently, we can not prove a general result of propagation of chaos if we do not take into account the diffusion coefficient  $\sqrt{\epsilon}$ .*

**Definition 3.5.** Set a continuous function  $f$  from  $\mathbb{R}$  to  $\mathbb{R}$  and  $\mathcal{X} \in \mathbb{R}^N$  with  $N \in \mathbb{N}^*$ . We define:  $f(\mathcal{X}) := \frac{1}{N} \sum_{i=1}^N f(X_i)$ . If  $\mu$  is a measure absolutely continuous with respect to the Lebesgue measure, we define  $f(\mu) := \int_{\mathbb{R}} f(x)\mu(x)dx$ .

We begin by providing a general theorem and after, we will derive three results from it.

**Theorem 3.6.** Set a law  $u_0$  such that  $u_t$  tends towards a measure  $\mu$  where  $u_t$  is the solution of (V) starting with  $u_0$ . We consider a sequence of iid random values with law  $u_0$ :  $(X_0^i)_{i \geq 1}$ . Set a  $C^\infty$ -function  $f$  such that  $|f(x) - f(y)| \leq C|x-y|(1+|x|+|y|)$  with  $C > 0$ . For all  $N \geq 1$ , we call  $\mathcal{X}_0^N := (X_0^1, \dots, X_0^N)$ . We introduce  $\mathcal{X}_t^N$  the solution of the random dynamical system (I) starting with  $\mathcal{X}_0^N$ . Then, for all  $\delta > 0$ , we have

$$\lim_{N \rightarrow +\infty} \mathbb{P} \{ |f(\mathcal{X}_s^N) - f(u_t)| < \delta ; \forall 0 \leq s \leq t \} = 1. \quad (3.3)$$

Furthermore, there exists  $T_\delta \geq 0$  such that for all  $t \geq 0$ , we have

$$\lim_{N \rightarrow +\infty} \mathbb{P} \{ |f(\mathcal{X}_s^N) - f(\mu)| < \delta ; \forall T_\delta \leq s \leq T_\delta + t \} = 1. \quad (3.4)$$

*Proof.* Let us prove the inequality (3.3). For all  $i \in \llbracket 1; N \rrbracket$ , set  $(\overline{X}_s^i)_{s \in [0; t]}$  the diffusion (IV) starting with  $X_0^i$ . By definition,  $\mathcal{L}(X_s^i) = u_s$ . We have immediately the following inequality:

$$\begin{aligned} & \mathbb{P} \left\{ \exists s \in [0; t] \mid |f(\mathcal{X}_s^N) - f(u_s)| \geq \delta \right\} = \mathbb{P} \left\{ \sup_{s \in [0; t]} |f(\mathcal{X}_s^N) - f(u_s)| \geq \delta \right\} \\ & \leq \mathbb{P} \left\{ \frac{1}{N} \sum_{i=1}^N \sup_{s \in [0; t]} |f(X_s^i) - f(\overline{X}_s^i)| \geq \frac{\delta}{2} \right\} \\ & + \mathbb{P} \left\{ \sup_{s \in [0; t]} \left| \frac{1}{N} \sum_{i=1}^N f(\overline{X}_s^i) - f(u_s) \right| \geq \frac{\delta}{2} \right\}. \end{aligned}$$

The second term tends towards 0 by applying the weak law of large numbers. Let us focus on the first one. We take  $f(x) := x$  and we apply Cauchy-Schwarz inequality and Proposition 3.1:

$$\begin{aligned} \mathbb{P} \left\{ \frac{1}{N} \sum_{i=1}^N \sup_{s \in [0; t]} |X_s^i - \overline{X}_s^i| \geq \frac{\delta}{2} \right\} & \leq \frac{2}{\delta} \max_{i \in \llbracket 1; N \rrbracket} \mathbb{E} \left\{ \sup_{s \in [0; t]} |X_s^i - \overline{X}_s^i| \right\} \\ & \leq \frac{2}{\delta} \sqrt{\frac{C}{N}} \exp \left[ \frac{Kt}{2} \right]. \end{aligned}$$

Since  $\mathbb{E} \left[ \left| \overline{X}_s^i \right| \right] \leq M_0$  defined in (1.2) for all  $i \in \llbracket 1; N \rrbracket$  and  $s \geq 0$ , we deduce that the same holds for  $X_s^i$  on any finite interval. Also, by using the same argument

than the one at the end of the proof of Proposition 3.1, we obtain

$$\max_{i \in \llbracket 1; N \rrbracket} \left\{ \mathbb{E} \left[ \sup_{s \in [0; t]} |X_s^i| \right] + \mathbb{E} \left[ \sup_{s \in [0; t]} |\overline{X}_s^i| \right] \right\} < \infty.$$

Let us take now a general  $f$ . The condition on  $f$  and  $C$ , the Cauchy-Schwarz inequality and Proposition 3.1 imply

$$\begin{aligned} \mathbb{P} \left\{ \frac{1}{N} \sum_{i=1}^N \sup_{[0; t]} |f(X_s^i) - f(\overline{X}_s^i)| \geq \frac{\delta}{2} \right\} &\leq \frac{2}{\delta} \max_{\llbracket 1; N \rrbracket} \mathbb{E} \left\{ \sup_{[0; t]} |f(X_s^i) - f(\overline{X}_s^i)| \right\} \\ &\leq \frac{2}{\delta} \sqrt{\frac{C}{N}} \exp \left[ \frac{Kt}{2} \right] \sqrt{1 + \max_{\llbracket 1; N \rrbracket} \left\{ \mathbb{E} \left[ \sup_{[0; t]} |X_s^i| \right] + \mathbb{E} \left[ \sup_{[0; t]} |\overline{X}_s^i| \right] \right\}} \rightarrow 0. \end{aligned}$$

In order to prove the second statement, it is sufficient to note that the tightness of the family  $(u_t)_{t \in \mathbb{R}_+}$  and the convergence of  $u_t$  towards  $\mu$  implies the convergence of  $f(u_t)$  towards  $f(\mu)$  so for all  $\delta > 0$ , there exists  $T_\delta \geq 0$  such that  $|f(u_t) - f(\mu)| \leq \frac{\delta}{2}$  for all  $t \geq T_\delta$  then we apply the first statement with  $\frac{\delta}{2}$ .  $\square$

The time  $T_\delta$  is deterministic and linked to the rate of convergence towards the stationary measure  $\mu$  so it depends on  $\epsilon$ .

With this general theorem, we will obtain three corollaries. The first one states that the mean-field system is prisoner of a ball.

**Corollary 3.7.** *Set  $\epsilon \geq \epsilon_c(\alpha)$  and a law  $\mu_0$  which satisfies (AC), (FM) and (FE). We consider a sequence of iid random values with law  $\mu_0$ :  $(X_0^i)_{i \geq 1}$ . For all  $N \geq 1$ , we call  $\mathcal{X}_0^N := (X_0^1, \dots, X_0^N)$ . We introduce  $\mathcal{X}_t^N$  the solution of the random dynamical system (I) starting with  $\mathcal{X}_0^N$ . Then, for all  $r > \sqrt{\text{Var}(u_\epsilon^0)}$ , there exists  $T_r \geq 0$  such that for all  $t \geq 0$ , we have*

$$\lim_{N \rightarrow +\infty} \mathbb{P} \left\{ \mathcal{X}_s^N \in \mathbb{B}_r^N(\overline{0}) ; \forall T_r \leq s \leq t + T_r \right\} = 1.$$

*Proof.* Theorem 2.1 in [Tug10b] states that  $\mu_t$  converges towards a stationary measure  $u_\epsilon$  when  $t$  tends to  $\infty$ . Since  $\epsilon \geq \epsilon_c(\alpha)$ , we know by Theorem 2.1 in [Tug11] that (IV) admits a unique stationary measure and it is  $u_\epsilon^0$  so  $\mu_t$  converges towards  $u_\epsilon^0$ . We conclude by applying Theorem 3.6 (more precisely we use the limit (3.4)) with  $\delta := r^2 - \text{Var}(u_\epsilon^0)$  and  $f(x) := x^2$ .  $\square$

We can not find smaller radius since  $f(u_\epsilon^0) = \text{Var}(u_\epsilon^0)$ .

**Remark 3.8.** *We can relax the assumption  $\epsilon \geq \epsilon_c(\alpha)$  if  $\mu_0$  is symmetric. Moreover, Theorem 3.1 in [HT09] implies that  $\text{Var}(u_\epsilon^0) = \frac{1}{4(\alpha-1)}\epsilon + o(\epsilon)$  if  $\alpha > 1$  and Theorem 4.5 in [HT09] implies that  $\text{Var}(u_\epsilon^0) = 1 - \alpha - \frac{\epsilon}{2(1-\alpha)} + o(\epsilon)$  if  $\alpha < 1$ . A simple computation provides also  $\text{Var}(u_\epsilon^0) = \frac{\Gamma(\frac{3}{4})}{\Gamma(\frac{1}{4})} \sqrt{\epsilon}$  if  $\alpha = 1$ .*

The second corollary provides sufficient condition for forbidding to cross the hyperplan  $\{\mathcal{X} \in \mathbb{R}^N \mid \sum_{i=1}^N X_i = 0\}$ .

**Corollary 3.9.** *Set  $\epsilon < \epsilon_c(\alpha)$  small enough and a law  $\mu_0$  which satisfies (AC), (FM) and (FE). Let us assume that  $\Upsilon_\epsilon(\mu_0) < \inf_{\{u \mid \int_{\mathbb{R}} xu(x)dx=0\}} \Upsilon_\epsilon(u)$  and  $\mathbb{E}(\mu_0) > 0$ . We consider a sequence of iid random values with law  $\mu_0$ :  $(X_0^i)_{i \geq 1}$ . For all  $N \geq 1$ , we call  $\mathcal{X}_0^N := (X_0^1, \dots, X_0^N)$ . We introduce  $\mathcal{X}_t^N$  the solution of the random dynamical system (I) starting with  $\mathcal{X}_0^N$ . Then for all  $t \geq 0$ , we have:*

$$\lim_{N \rightarrow +\infty} \mathbb{P} \{ \mathcal{X}_s^N \in \mathbb{E}_+^N ; \forall 0 \leq s \leq t \} = 1.$$

*Proof.* We recall that the free-energy is nonincreasing. Consequently,  $\Upsilon_\epsilon(u_s) < \inf_{\{u \mid \int_{\mathbb{R}} xu(x)dx=0\}} \Upsilon_\epsilon(u)$  for all  $s \in [0; t]$ . This implies  $\int_{\mathbb{R}} xu_s(x)dx > 0$  for all  $s \in [0; t]$ . We conclude by applying Theorem 3.6, more precisely the limit (3.3) with  $f(x) := x$  and  $\delta := \inf_{s \in [0; t]} \int_{\mathbb{R}} xu_s(x)dx > 0$ .  $\square$

Finally, the last corollary stresses the fact that the steady states do not correspond to the wells of  $\Upsilon^N$ .

**Corollary 3.10.** *Set  $p \in ]0; 1[$  with  $p \neq \frac{1}{2}$  and a law  $\mu_0$  which satisfies (AC), (FM) and (FE). We assume that  $\mu_0$  has the signature  $(p, 1-p)$ . We consider a sequence of iid random values with law  $\mu_0$ :  $(X_0^i)_{i \geq 1}$ . For all  $N \geq 1$ , we call  $\mathcal{X}_0^N := (X_0^1, \dots, X_0^N)$ . We introduce  $\mathcal{X}_t^N$  the solution of the random dynamical system (I) starting with  $\mathcal{X}_0^N$ . Set  $\kappa > 0$ . There exists  $\epsilon_0 > 0$  such that for all  $\epsilon \in ]0; \epsilon_0[$ , there exists  $T_\kappa \geq 0$  such that for all  $t > 0$ , we have:*

$$\lim_{N \rightarrow +\infty} \mathbb{P} \left[ \mathcal{X}_s \in \bigcup_{\rho=0}^{\kappa} \left( \mathbb{S}_\rho^N \cup \mathbb{S}_{1-\rho}^N \cup \mathbb{S}_{\frac{1}{2}+\rho}^N \cup \mathbb{S}_{\frac{1}{2}-\rho}^N \right), \forall T_\kappa \leq s \leq T_\kappa + t \right] = 1.$$

*Proof.* The law  $u_t$  converges towards  $u_\epsilon^0$ ,  $u_\epsilon^+$  or  $u_\epsilon^-$  according to Theorem 2.1 in [Tug10b]. Since  $u_\epsilon^0$  is symmetric,  $\int_{\mathbb{R}} \mathbb{1}_{]0; +\infty[}(x) u_\epsilon^0(x) dx = \frac{1}{2}$ . Theorem 2.4 in [HT10b] proves that  $u_\epsilon^\pm$  converges towards  $\delta_{\pm 1}$  when  $\epsilon$  tends to 0. So, for  $\epsilon$  small enough, we have  $\int_{\mathbb{R}} \mathbb{1}_{]0; +\infty[}(x) u_\epsilon^+(x) dx \geq 1 - \frac{\kappa}{3}$  and  $\int_{\mathbb{R}} \mathbb{1}_{]0; +\infty[}(x) u_\epsilon^-(x) dx \leq \frac{\kappa}{3}$ . We apply Theorem 3.6, more precisely (3.4) with  $\delta := \frac{\kappa}{3}$  and with

$$f(x) := \mathbb{1}_{] \frac{\eta}{3}; +\infty[}(x) + \mathbb{1}_{[0; \frac{\eta}{3}]}(x) Z^{-1} \int_0^x \exp \left[ -\frac{1}{y^2} - \frac{1}{(y - \frac{\kappa}{3})^2} \right] dy$$

where  $Z := \int_0^{\eta/3} \exp \left[ -\frac{1}{y^2} - \frac{1}{(y - \kappa/3)^2} \right] dy$  is such that  $f(\frac{\eta}{3}) = 1$ . We take  $\eta$  sufficiently small for having  $|f(u_\epsilon^0) - \frac{1}{2}| \leq \frac{\kappa}{3}$ ,  $|f(u_\epsilon^\pm) - (\pm 1)| \leq \frac{\kappa}{3}$ . The proof is achieved by applying Theorem 3.6 with  $f$  and  $\delta := \frac{\kappa}{3}$ .  $\square$

This theorem proves that the wells with signature  $(p, 1-p)$  are not stable and even if it is possible to have  $2^N(1-o(1))$  wells, these points do not intervene in the dynamic that achieves to prove that the meta-potential is not sufficient for understanding the behavior of the mean-field system (I).

## References

- [BAZ99] G. Ben Arous and O. Zeitouni. Increasing propagation of chaos for mean field models. *Ann. Inst. H. Poincaré Probab. Statist.*, 35(1):85–102, 1999.
- [BBI09] Alessandra Bianchi, Anton Bovier, and Dmitry Ioffe. Sharp asymptotics for metastability in the random field Curie-Weiss model. *Electron. J. Probab.*, 14:no. 53, 1541–1603, 2009.
- [BCCP98] D. Benedetto, E. Caglioti, J. A. Carrillo, and M. Pulvirenti. A non-Maxwellian steady distribution for one-dimensional granular media. *J. Statist. Phys.*, 91(5-6):979–990, 1998.
- [BFG07] Nils Berglund, Bastien Fernandez, and Barbara Gentz. Metastability in interacting nonlinear stochastic differential equations. II. Large- $N$  behaviour. *Nonlinearity*, 20(11):2583–2614, 2007.
- [BGV07] François Bolley, Arnaud Guillin, and Cédric Villani. Quantitative concentration inequalities for empirical measures on non-compact spaces. *Probab. Theory Related Fields*, 137(3-4):541–593, 2007.
- [BRTV98] S. Benachour, B. Roynette, D. Talay, and P. Vallois. Nonlinear self-stabilizing processes. I. Existence, invariant probability, propagation of chaos. *Stochastic Process. Appl.*, 75(2):173–201, 1998.
- [BRV98] S. Benachour, B. Roynette, and P. Vallois. Nonlinear self-stabilizing processes. II. Convergence to invariant probability. *Stochastic Process. Appl.*, 75(2):203–224, 1998.
- [CDPS10] Francesca Collet, Paolo Dai Pra, and Elena Sartori. A simple mean field model for social interactions: dynamics, fluctuations, criticality. *J. Stat. Phys.*, 139(5):820–858, 2010.
- [CGM08] P. Cattiaux, A. Guillin, and F. Malrieu. Probabilistic approach for granular media equations in the non-uniformly convex case. *Probab. Theory Related Fields*, 140(1-2):19–40, 2008.
- [CMV03] José A. Carrillo, Robert J. McCann, and Cédric Villani. Kinetic equilibration rates for granular media and related equations: entropy dissipation and mass transportation estimates. *Rev. Mat. Iberoamericana*, 19(3):971–1018, 2003.
- [CX10] Dan Crisan and Jie Xiong. Approximate McKean-Vlasov representations for a class of SPDEs. *Stochastics*, 82(1-3):53–68, 2010.
- [Der03] Azzouz Dermoune. Propagation and conditional propagation of chaos for pressureless gas equations. *Probab. Theory Related Fields*, 126(4):459–476, 2003.
- [DG87] D.A Dawson and J Gärtner. Large deviations from the McKean-Vlasov limit for weakly interacting diffusions. *Stochastics*, 20(4):247–308, 1987.
- [DPdH96] Paolo Dai Pra and Frank den Hollander. McKean-Vlasov limit for interacting random processes in random media. *J. Statist. Phys.*, 84(3-4):735–772, 1996.
- [DZ10] Amir Dembo and Ofer Zeitouni. *Large deviations techniques and applications*, volume 38 of *Stochastic Modelling and Applied Probability*. Springer-Verlag, Berlin, 2010. Corrected reprint of the second (1998) edition.

- [FW98] M. I. Freidlin and A. D. Wentzell. *Random perturbations of dynamical systems*, volume 260 of *Grundlehren der Mathematischen Wissenschaften [Fundamental Principles of Mathematical Sciences]*. Springer-Verlag, New York, second edition, 1998. Translated from the 1979 Russian original by Joseph Szücs.
- [Gra90] Carl Graham. Nonlinear limit for a system of diffusing particles which alternate between two states. *Appl. Math. Optim.*, 22(1):75–90, 1990.
- [Gra92] Carl Graham. McKean-Vlasov Itô-Skorohod equations, and nonlinear diffusions with discrete jump sets. *Stochastic Process. Appl.*, 40(1):69–82, 1992.
- [HIP08] Samuel Herrmann, Peter Imkeller, and Dierk Peithmann. Large deviations and a Kramers’ type law for self-stabilizing diffusions. *Ann. Appl. Probab.*, 18(4):1379–1423, 2008.
- [HT09] S. Herrmann and J. Tugaut. Self-stabilizing processes: uniqueness problem for stationary measures and convergence rate in the small noise limit. *Prépublications de l’Institut Élie Cartan*, 2009.
- [HT10a] S. Herrmann and J. Tugaut. Non-uniqueness of stationary measures for self-stabilizing processes. *Stochastic Process. Appl.*, 120(7):1215–1246, 2010.
- [HT10b] S. Herrmann and J. Tugaut. Stationary measures for self-stabilizing processes: asymptotic analysis in the small noise limit. *Electron. J. Probab.*, 15:2087–2116, 2010.
- [JM08] Benjamin Jourdain and Florent Malrieu. Propagation of chaos and Poincaré inequalities for a system of particles interacting through their CDF. *Ann. Appl. Probab.*, 18(5):1706–1736, 2008.
- [Mal01] F. Malrieu. Logarithmic Sobolev inequalities for some nonlinear PDE’s. *Stochastic Process. Appl.*, 95(1):109–132, 2001.
- [Mal03] Florent Malrieu. Convergence to equilibrium for granular media equations and their Euler schemes. *Ann. Appl. Probab.*, 13(2):540–560, 2003.
- [McK67] H. P. McKean, Jr. Propagation of chaos for a class of non-linear parabolic equations. In *Stochastic Differential Equations (Lecture Series in Differential Equations, Session 7, Catholic Univ., 1967)*, pages 41–57. Air Force Office Sci. Res., Arlington, Va., 1967.
- [Mél96] Sylvie Méléard. Asymptotic behaviour of some interacting particle systems; McKean-Vlasov and Boltzmann models. In *Probabilistic models for nonlinear partial differential equations (Montecatini Terme, 1995)*, volume 1627 of *Lecture Notes in Math.*, pages 42–95. Springer, Berlin, 1996.
- [MP98] P. Mathieu and P. Picco. Metastability and convergence to equilibrium for the random field Curie-Weiss model. *J. Statist. Phys.*, 91(3-4):679–732, 1998.
- [Szn91] Alain-Sol Sznitman. Topics in propagation of chaos. In *École d’Été de Probabilités de Saint-Flour XIX—1989*, volume 1464 of *Lecture Notes in Math.*, pages 165–251. Springer, Berlin, 1991.
- [Tug10a] J. Tugaut. *Processus autostabilisants dans un paysage multi-puits*. PhD thesis, Université Henri Poincaré, Nancy, 2010.

- [Tug10b] J. Tugaut. Convergence to the equilibria for self-stabilizing processes in double-well landscape. *Preprint, Bielefeld Universität*, 2010.
- [Tug11] J. Tugaut. Phase transitions of McKean-Vlasov processes in symmetric and asymmetric multi-wells landscape. *Preprint, Bielefeld Universität*, 2011.
- [Ver06] A. Yu. Veretennikov. On ergodic measures for McKean-Vlasov stochastic equations. In *Monte Carlo and quasi-Monte Carlo methods 2004*, pages 471–486. Springer, Berlin, 2006.