

# On the Smoluchowski-Kramers approximation in the presence of a varying magnetic field

Sandra Cerrai\*

University of Maryland, College Park, USA

Jan Wehr†

University of Arizona, Tucson, USA

Yichun Zhu

University of Maryland, College Park, USA

## Abstract

We study the small mass limit for the equation that describes the planar motion of a charged particle of a small mass  $\mu$  in a force field that has a deterministic as well as a stochastic component, combined with a magnetic field. We regularize the problem by adding a small friction of intensity  $\epsilon > 0$ . We show that for all small but fixed frictions the small mass limit for  $q_{\mu,\epsilon}$  gives the solution  $q_\epsilon$  to a stochastic first order equation, where a noise-induced drift term is created. Then, by using a generalization of the classical averaging theorem for Hamiltonian systems by Freidlin and Wentzell, we take the limit of the slow component of the motion  $q_\epsilon$  and we prove that it converges weakly to a Markov process on the graph obtained by identifying all points in the same connected components of the level sets of the intensity function of the magnetic field.

## 1 Introduction

We are dealing with the planar motion of a charged particle of a small mass  $\mu$  in a force field that has a deterministic as well as a stochastic component combined with a magnetic field

$$\begin{cases} \mu \ddot{q}_\mu(t) = b(q_\mu(t)) - \lambda(q_\mu(t))A\dot{q}_\mu(t) + \sigma(q_\mu(t))\dot{w}_t, \\ q_\mu(0) = q \in \mathbb{R}^2, \quad \dot{q}_\mu(0) = p \in \mathbb{R}^2. \end{cases} \quad (1.1)$$

Here  $b$  is a vector field in  $\mathbb{R}^2$ ,  $\sigma$  is  $2 \times 2$ -matrix valued mapping defined on  $\mathbb{R}^2$  and  $w(t)$  is a standard two-dimensional Brownian motion. Moreover,  $\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}$  is some mapping, such that  $\lambda(x) \geq \lambda_0 > 0$ , for every  $x \in \mathbb{R}^2$ , and

$$A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}.$$

We are here interested in understanding the limiting behavior of the solution  $q_\mu$  to equation (1.1), as the mass  $\mu$  vanishes. This is the so called Smoluchowski-Kramers approximation.

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It is well known (see [9] for all details) that when the variable magnetic field considered in the present paper is replaced by a constant friction (that is  $\lambda$  is constant and the matrix  $A$  coincides with the identity matrix), then  $q_\mu(t)$  can be approximated with the solution of the first order equation

$$dq(t) = b(q(t)) dt + \sigma(q(t)) dw(t), \quad q(0) = q. \quad (1.2)$$

More precisely, for every fixed  $T > 0$

$$\lim_{\mu \rightarrow 0} \mathbb{E} \max_{t \in [0, T]} |q_\mu(t) - q(t)|^2 = 0. \quad (1.3)$$

Notice that here the case of an arbitrary number of degrees of freedom can be covered. The same result can be obtained also if  $\lambda$  is still constant, but  $A$  is a more general matrix, whose eigenvalues have strictly positive real part, with the limiting equation (1.2) replaced by

$$dq(t) = A^{-1}b(q(t)) dt + A^{-1}\sigma(q(t)) dw(t), \quad q(0) = q. \quad (1.4)$$

The case of non constant friction has been widely studied recently (see [12] and [13] for example). They have considered the following system

$$\begin{cases} \mu \ddot{q}_\mu(t) = b(q_\mu(t)) - \gamma(q_\mu(t)) \dot{q}_\mu(t) + \sigma(q_\mu(t)) \dot{w}_t, \\ q_\mu(0) = q \in \mathbb{R}^k, \quad \dot{q}_\mu(0) = p \in \mathbb{R}^k, \end{cases} \quad (1.5)$$

for some  $h$ -dimensional Brownian motion  $w(t)$ . They have assumed that the coefficients  $b : \mathbb{R}^k \rightarrow \mathbb{R}^k$ ,  $\gamma : \mathbb{R}^k \rightarrow \mathbb{R}^{k \times k}$  and  $\sigma : \mathbb{R}^k \rightarrow \mathbb{R}^{h \times k}$  are smooth and uniformly bounded and the smallest eigenvalue  $\lambda_1(q)$  of the symmetric matrix  $\gamma(q) + \gamma^*(q)$  is strictly positive, uniformly with respect to  $q \in \mathbb{R}^k$ . Namely

$$\inf_{q \in \mathbb{R}^k} \lambda_1(q) =: \bar{\lambda} > 0.$$

They have proved that limit (1.3) is still valid, but now  $q(t)$  is the solution of the modified equation

$$dq(t) = [\gamma^{-1}(q(t))b(q(t)) + S(q(t))] dt + \gamma^{-1}(q(t))\sigma(q(t))dw(t), \quad q(0) = q, \quad (1.6)$$

where  $S(q)$  is the noise-induced drift whose  $j$ -th component equals

$$S_j(q) = \sum_{i,l=1}^k \frac{\partial}{\partial q_i} (\gamma^{-1})_{jl}(q) J_{li}(q), \quad j = 1, \dots, k,$$

where  $J$  is the matrix-valued function solving the Lyapunov equation

$$J(q)\gamma^*(q) + \gamma(q)J(q) = \sigma(q)\sigma^*(q), \quad q \in \mathbb{R}^k.$$

In [4], the case of a particle subject to a constant strength magnetic field orthogonal to the plane where the particle moves has been considered. In this case, the motion of the particle is governed by equation (1.1), with  $\lambda(q) \equiv \bar{\lambda}$ , for every  $q \in \mathbb{R}^2$  (for semplicity of notation in what follows we shall take  $\bar{\lambda} = 1$ ). In particular, since the eigenvalues of  $A$  are purely imaginary, the methods and results described above are not valid anymore.

It is not difficult to check that if the stochastic term in (1.1) is replaced by a continuous function, then  $q_\mu$  converges uniformly in  $[0, T]$  to the solution of (1.4). But if such continuous function is replaced by white noise, then there is no more convergence of  $q_\mu$  to the solution of (1.4), as  $\mu \downarrow 0$ . Actually, while

$$\lim_{\mu \rightarrow 0} \int_0^t \sin \frac{s}{\mu} \varphi(s) ds = 0,$$

for every continuous function, when  $w(t)$  is a Brownian motion we have

$$\text{Var} \left( \int_0^t \sin \frac{s}{\mu} dw(s) \right) = \int_0^t \sin^2 \frac{s}{\mu} ds \rightarrow \frac{t}{2}, \quad \text{as } \mu \downarrow 0,$$

so that

$$\lim_{\mu \rightarrow 0} \int_0^t \sin \frac{s}{\mu} dw(s) \neq 0.$$

Because of this, in [4] the problem has been regularized, so that a suitable counterpart of the Smoluchowski-Kramers approximation has been proved. The first regularization consisted in introducing in equation (1.1) a small friction proportional to the velocity. Namely, the following equation has been considered

$$\begin{cases} \mu \ddot{q}_{\mu,\epsilon}(t) = b(q_{\mu,\epsilon}(t)) - A_\epsilon \dot{q}_{\mu,\epsilon}(t) + \sigma(q_{\mu,\epsilon}(t)) \dot{w}(t), \\ q_{\mu,\epsilon}(0) = q \in \mathbb{R}^2, \quad \dot{q}_{\mu,\epsilon}(0) = p \in \mathbb{R}^2, \end{cases}$$

where  $A_\epsilon = A + \epsilon I$  and  $\epsilon > 0$  is a small parameter. It has been shown that for any  $T > 0$

$$\lim_{\mu \rightarrow 0} \mathbb{E} \max_{t \in [0, T]} |q_{\mu,\epsilon}(t) - q_\epsilon(t)|^2 = 0, \quad (1.7)$$

where  $q_\epsilon(t)$  is the solution of the problem

$$dq(t) = A_\epsilon^{-1} b(q(t)) dt + A_\epsilon^{-1} \sigma(q(t)) dw(t), \quad q(0) = q.$$

Next, it has been shown that

$$\lim_{\epsilon \rightarrow 0} \mathbb{E} \max_{t \in [0, T]} |q_\epsilon(t) - q(t)|^2 = 0,$$

where  $q(t)$  is the solution of the problem

$$dq(t) = -A b(q(t)) dt - A \sigma(q(t)) dw(t), \quad q(0) = q. \quad (1.8)$$

Another approach to regularization (see also [15] for the case of non constant magnetic field) used the fact that the white noise  $\dot{w}(t)$  can be considered as an idealization of an isotropic  $\delta$ -correlated smooth mean-zero Gaussian process  $\dot{w}^\delta(t)$ , with  $0 < \delta \ll 1$ , which converges to the standard white noise  $w(t)$ , as  $\delta \downarrow 0$ . In this case, it has been proven that if  $q_{\mu,\delta}(t)$  is the solution of equation (1.1), with  $\dot{w}(t)$  replaced by  $\dot{w}^\delta(t)$ , then

$$\lim_{\mu \rightarrow 0} \mathbb{E} \max_{t \in [0, T]} |q_{\mu,\delta}(t) - q_\delta(t)| = 0,$$

where  $q_\delta(t)$  solves the equation

$$\dot{q}(t) = -Ab(q(t)) - A\sigma(q(t))\dot{w}^\delta(t), \quad q(0) = q.$$

Next, by taking the limit as  $\delta \downarrow 0$ , it has been proven that  $q_\delta(t)$  converges to the solution  $\hat{q}(t)$  of the problem

$$d\hat{q}(t) = -Ab(\hat{q}(t))dt - A\sigma(\hat{q}(t)) \circ dw(t), \quad \hat{q}(0) = q,$$

where the stochastic term has to be interpreted in Stratonovich sense.

In the present paper we are interested in the small mass limit in presence of a non-constant magnetic field. To this purpose we proceed by adding a small constant friction and we consider the regularized equation

$$\begin{cases} \mu \ddot{q}_{\mu,\epsilon}(t) = b(q_{\mu,\epsilon}(t)) - [\lambda(q_{\mu,\epsilon}(t))A + \epsilon I] \dot{q}_{\mu,\epsilon}(t) + \sigma(q_{\mu,\epsilon}(t)) \dot{w}_t, \\ q_{\mu,\epsilon}(0) = q \in \mathbb{R}^2, \quad \dot{q}_{\mu,\epsilon}(0) = p \in \mathbb{R}^2. \end{cases} \quad (1.9)$$

We show that under suitable conditions on the coefficients  $b$ ,  $\sigma$  and  $\lambda$ , the problem above is well posed in  $L^k(\Omega; C([0, T]; \mathbb{R}^2))$ , for every  $T > 0$  and  $k \geq 1$ .

For every fixed  $\epsilon > 0$ , equation (1.9) is of the same type as those considered in [12] and [13], so that we can take the small mass limit as  $\mu$  goes to zero and we obtain that for every  $\epsilon > 0$

$$\lim_{\mu \rightarrow 0} \mathbb{E} \sup_{t \in [0, T]} |q_{\mu,\epsilon}(t) - q_\epsilon(t)| = 0,$$

where  $q_\epsilon$  is the solution of the problem

$$\begin{cases} dq_\epsilon(t) = \left[ (\lambda(q_\epsilon(t))A + \epsilon I)^{-1} b(q_\epsilon(t)) + S_\epsilon(q_\epsilon(t)) \right] dt + (\lambda(q_\epsilon(t))A + \epsilon I)^{-1} \sigma(q_\epsilon(t)) dw(t), \\ q_\epsilon(0) = q. \end{cases}$$

After some computations, it turns out that  $q_\epsilon$  solves the equation

$$\begin{aligned} dq_\epsilon(q) &= \frac{1}{\epsilon} \gamma(q_\epsilon(t)) \nabla^\perp \lambda(q_\epsilon(t)) dt + B(q_\epsilon(t)) dt + \Sigma(q_\epsilon(t)) dw(t), \\ &+ \epsilon [B_\epsilon(q_\epsilon(t)) dt + \Sigma_\epsilon(q_\epsilon(t)) dw(t)], \quad q_\epsilon(0) = q, \end{aligned}$$

for some mappings  $\gamma : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $B, B_\epsilon : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  and  $\Sigma, \Sigma_\epsilon : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  that are explicitly given. This means that the motion of  $q_\epsilon$  is made of a fast component on the level sets of  $\lambda$  and a slow transversal motion. Thus, by using a suitable generalization of the classical result of Freidlin and Wentzell on averaging for Hamiltonian systems (see [11, Chapter 8] and [24]), we prove that the projection of  $q_\epsilon$  over the graph  $\Gamma$ , obtained by identifying all points on the same connected component of each level set of  $\lambda$ , converges to a suitable Markov process  $Y$ , whose generator is explicitly given.

## 2 Well-posedness of the regularized problem

As we mentioned in the Introduction, we are dealing here with the following equation

$$\begin{cases} \mu \ddot{q}_\mu(t) = b(q_\mu(t)) - \lambda(q_\mu(t))A\dot{q}_\mu(t) + \sigma(q_\mu(t))\dot{w}_t, \\ q_\mu(0) = q \in \mathbb{R}^2, \quad \dot{q}_\mu(0) = p \in \mathbb{R}^2, \end{cases} \quad (2.1)$$

where  $\mu$  is a small positive constant and  $w(t)$  is a standard Brownian motion in  $\mathbb{R}^2$ .

In this section, we shall assume that the coefficients in the equation above satisfy the following conditions. In fact, in Section 4 we will impose a more restrictive growth condition on  $\lambda$ .

**Hypothesis 1.** 1. The mappings  $b : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  and  $\sigma : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  are Lipschitz-continuous.

2. The mapping  $\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}$  is locally Lipschitz-continuous and there exist  $\gamma \geq 0$  and  $c > 0$  such that

$$|\lambda(q)| \leq c(1 + |q|^\gamma), \quad \lambda \in \mathbb{R}^2. \quad (2.2)$$

Moreover

$$\inf_{q \in \mathbb{R}^2} \lambda(q) =: \lambda_0 > 0.$$

Next, for every  $\epsilon \geq 0$  we introduce the regularized problem

$$\begin{cases} \mu \ddot{q}_{\mu,\epsilon}(t) = b(q_{\mu,\epsilon}(t)) - \Lambda_\epsilon(q_{\mu,\epsilon}(t))\dot{q}_{\mu,\epsilon}(t) + \sigma(q_{\mu,\epsilon}(t))\dot{w}_t, \\ q_{\mu,\epsilon}(0) = q \in \mathbb{R}^2, \quad \dot{q}_{\mu,\epsilon}(0) = p \in \mathbb{R}^2, \end{cases} \quad (2.3)$$

where

$$\Lambda_\epsilon(q) = \lambda(q)A + \epsilon I = \begin{pmatrix} \epsilon & \lambda(q) \\ -\lambda(q) & \epsilon \end{pmatrix}, \quad q \in \mathbb{R}^2.$$

Notice that for every  $\epsilon > 0$  the matrix  $\Lambda_\epsilon(q)$  is uniformly non-degenerate, as

$$\langle \Lambda_\epsilon(q)p, p \rangle = \epsilon |p|^2. \quad (2.4)$$

Moreover, when  $\epsilon = 0$ , equation (2.3) coincides with equation (2.1).

**Theorem 2.1.** Under Hypothesis 1, for every  $\mu > 0$  and  $\epsilon \geq 0$  and for every  $T > 0$  and  $k \geq 1$ , equation (2.3) admits a unique adapted solution  $q_{\mu,\epsilon} \in L^k(\Omega; C([0, T]; \mathbb{R}^2))$ .

*Proof.* For every  $q, p \in \mathbb{R}^2$  and  $n \in \mathbb{N}$ , we define

$$\beta_n(p) = \begin{cases} p, & \text{if } |p| \leq n, \\ np/|p|, & \text{if } |p| \geq n, \end{cases}$$

and

$$\Lambda_{\epsilon,n}(q) = \lambda_n(q)A + \epsilon I, \quad \text{where} \quad \lambda_n(q) = \begin{cases} \lambda(q), & \text{if } |q| \leq n, \\ \lambda((n+1)q/|q|), & \text{if } |q| \geq n+1, \end{cases}.$$

Notice that  $\lambda_n : \mathbb{R}^2 \rightarrow \mathbb{R}$  is Lipschitz-continuous and

$$|\lambda_n(q)| \leq c(1 + |q|^\gamma), \quad |\beta_n(p)| \leq |p|, \quad (2.5)$$

for some constant  $c$  independent of  $n$ . Moreover, since  $\langle A\beta_n(p), p \rangle = 0$ , and  $\langle \beta_n(p), p \rangle \leq |p|^2$ , for every  $p \in \mathbb{R}^2$  and  $n \in \mathbb{N}$ , we have that

$$\langle \Lambda_{\epsilon,n}(q)\beta_n(p), p \rangle = \epsilon |p|^2, \quad (2.6)$$

for every  $p, q \in \mathbb{R}^2$ ,  $n \in \mathbb{N}$  and  $\epsilon > 0$ .

With these notations, we introduce the problem

$$\begin{cases} \mu \ddot{q}_{\mu,\epsilon}^n(t) = b(q_{\mu,\epsilon}^n(t)) - \Lambda_{\epsilon,n}(q_{\mu,\epsilon}^n(t))\beta_n(\dot{q}_{\mu,\epsilon}^n(t)) + \sigma(q_{\mu,\epsilon}^n(t)) \dot{w}_t, \\ q_{\mu,\epsilon}^n(0) = q \in \mathbb{R}^2, \quad \dot{q}_{\mu,\epsilon}^n(0) = p \in \mathbb{R}^2, \end{cases}$$

which can be rewritten as

$$\begin{cases} dq_{\mu,\epsilon}^n(t) = p_{\mu,\epsilon}^n(t) dt, \quad q_{\mu,\epsilon}^n(0) = q \\ \mu dp_{\mu,\epsilon}^n(t) = [b(q_{\mu,\epsilon}^n(t)) - \Lambda_{\epsilon,n}(q_{\mu,\epsilon}^n(t))\beta_n(p_{\mu,\epsilon}^n(t))] + \sigma(q_{\mu,\epsilon}^n(t)) dw(t), \quad p_{\mu,\epsilon}^n(t0) = p. \end{cases} \quad (2.7)$$

It is immediate to check that, for every fixed  $n \in \mathbb{N}$  and  $\epsilon > 0$ , the mapping

$$(q, p) \in \mathbb{R}^2 \times \mathbb{R}^2 \mapsto \Lambda_{\epsilon,n}(q)\beta_n(p) \in \mathbb{R}^2,$$

is Lipschitz-continuous, so that equation (2.7) admits a unique adapted solution  $(q_{\mu,\epsilon}^n, p_{\mu,\epsilon}^n) \in L^p(\Omega; C^1([0, T]; \mathbb{R}^2) \times C([0, T]; \mathbb{R}^2))$ .

Now, if we apply Itô's formula to the function  $\Phi(q, p) = |q|^{2k} + |p|^{2k}$ , for  $k \geq 2$ , we obtain

$$\begin{aligned} |q_{\mu,\epsilon}^n(t)|^{2k} + |p_{\mu,\epsilon}^n(t)|^{2k} &= |q|^{2k} + |p|^{2k} + k \int_0^t |q_{\mu,\epsilon}^n(s)|^{2k-2} \langle q_{\mu,\epsilon}^n(s), p_{\mu,\epsilon}^n(s) \rangle ds \\ &+ \frac{k}{\mu} \int_0^t |p_{\mu,\epsilon}^n(s)|^{2k-2} \langle p_{\mu,\epsilon}^n(s), b(q_{\mu,\epsilon}^n(s) - \Lambda_{\epsilon,n}(q_{\mu,\epsilon}^n(s))\beta_n(p_{\mu,\epsilon}^n(s))) \rangle ds \\ &+ \frac{k}{2\mu^2} \int_0^t |p_{\mu,\epsilon}^n(s)|^{2k-2} \text{Tr} [\sigma \sigma^*(q_{\mu,\epsilon}^n(s))] ds + \frac{k(k-1)}{2\mu^2} \int_0^t |p_{\mu,\epsilon}^n(s)|^{2k-4} |\sigma(q_{\mu,\epsilon}^n(s)) p_{\mu,\epsilon}^n(s)|^2 ds \\ &+ \frac{k}{\mu} \int_0^t |p_{\mu,\epsilon}^n(s)|^{2k-2} \langle p_{\mu,\epsilon}^n(s), \sigma(q_{\mu,\epsilon}^n(s)) dw(s) \rangle. \end{aligned}$$

Therefore, thanks to (2.6) and to the Young inequality, we have that for every  $\epsilon > 0$

$$\begin{aligned} |q_{\mu,\epsilon}^n(t)|^{2k} + |p_{\mu,\epsilon}^n(t)|^{2k} &\leq |q|^{2k} + |p|^{2k} + c_{k,\mu} \int_0^t [|q_{\mu,\epsilon}^n(s)|^{2k} + |p_{\mu,\epsilon}^n(s)|^{2k}] ds \\ &+ \frac{k}{\mu} \int_0^t |p_{\mu,\epsilon}^n(s)|^{2k-2} \langle p_{\mu,\epsilon}^n(s), \sigma(q_{\mu,\epsilon}^n(s)) dw(s) \rangle. \end{aligned}$$

After we take expectation in both sides, due to the Gronwall lemma we obtain

$$\mathbb{E}|q_{\mu,\epsilon}^n(t)|^{2k} + \mathbb{E}|p_{\mu,\epsilon}^n(t)|^{2k} \leq c_{k,\mu}(T) \left( 1 + |q|^{2k} + |p|^{2k} \right), \quad t \in [0, T]. \quad (2.8)$$

Therefore, since

$$q_{\mu,\epsilon}^n(t) = q + \int_0^t p_{\mu,\epsilon}^n(s) \, ds,$$

and

$$p_{\mu,\epsilon}^n(t) = p + \frac{1}{\mu} \int_0^t [b(q_{\mu,\epsilon}^n(s)) - \Lambda_{\epsilon,n}(q_{\mu,\epsilon}^n(s))\beta_n(p_{\mu,\epsilon}^n(s))] \, ds + \frac{1}{\mu} \int_0^t \sigma(q_{\mu,\epsilon}^n(s)) \, dw(s),$$

due to (2.5), from (2.8) we obtain

$$\sup_{n \in \mathbb{N}} \mathbb{E} \sup_{t \in [0, T]} \left( |q_{\mu,\epsilon}^n(t)|^{2k} + |p_{\mu,\epsilon}^n(t)|^{2k} \right) \leq c_{k,\mu}(T, |q|, |p|). \quad (2.9)$$

Now, for any  $n \in \mathbb{N}$  we define

$$\tau_n = \inf \{t \geq 0 : |q_{\mu,\epsilon}^n(t)| \vee |p_{\mu,\epsilon}^n(t)| \geq n\},$$

with the usual convention that  $\inf \emptyset = +\infty$ . Since

$$(q_{\mu,\epsilon}^n(t), p_{\mu,\epsilon}^n(t)) = (q_{\mu,\epsilon}^m(t), p_{\mu,\epsilon}^m(t)), \quad n < m, \quad t \leq \tau_n, \quad (2.10)$$

it follows that the sequence  $\{\tau_n\}_{n \in \mathbb{N}}$  is non-decreasing,  $\mathbb{P}$ -a.s., so that we can define

$$\tau = \lim_{n \rightarrow \infty} \tau_n.$$

Due to (2.9), for every fixed  $T > 0$  we have

$$\begin{aligned} & \mathbb{P} \left( \sup_{t \in [0, T]} |q_{\mu,\epsilon}^n(t)| \leq n, \sup_{t \in [0, T]} |p_{\mu,\epsilon}^n(t)| \leq n \right) \\ & \geq 1 - \mathbb{P} \left( \sup_{t \in [0, T]} |q_{\mu,\epsilon}^n(t)| > n \right) - \mathbb{P} \left( \sup_{t \in [0, T]} |p_{\mu,\epsilon}^n(t)| > n \right) \\ & \geq 1 - \frac{2c_{1,\mu}(T, |q|, |p|)}{n}. \end{aligned}$$

This implies that

$$\lim_{n \rightarrow \infty} \mathbb{P}(\tau_n > T) = 1,$$

and then, due to the arbitrariness of  $T$ , we conclude

$$\mathbb{P}(\tau = +\infty) = 1.$$

In particular, if we set

$$(q_{\mu,\epsilon}(t), p_{\mu,\epsilon}(t)) = (q_{\mu,\epsilon}^n(t \wedge \tau_n), p_{\mu,\epsilon}^n(t \wedge \tau_n)), \quad t \leq \tau,$$

due to (2.10) we can conclude that there exists a unique solution  $(q_{\mu,\epsilon}, p_{\mu,\epsilon})$  to problem (2.3), belonging to  $L^k(\Omega; C^1([0, T]; \mathbb{R}^2) \times C([0, T]; \mathbb{R}^2))$ , for every  $k \geq 1$  and  $T > 0$ .  $\square$

### 3 The Smoluchowski-Kramers approximation for the regularized problem

It is immediate to check that for every  $\epsilon > 0$  and  $q \in \mathbb{R}^2$ , the matrix  $\Lambda_\epsilon(q)$  is invertible and

$$\Lambda_\epsilon^{-1}(q) = \frac{1}{\lambda^2(q) + \epsilon^2} \begin{pmatrix} \epsilon & -\lambda(q) \\ \lambda(q) & \epsilon \end{pmatrix}. \quad (3.1)$$

Now, we introduce the vector field  $S^\epsilon(q)$ , whose  $j$ -th component is defined by

$$S_j^\epsilon(q) = \sum_{i,l=1}^2 \partial_i (\Lambda_\epsilon^{-1})_{jl}(q) J_{li}^\epsilon(q), \quad j = 1, 2, \quad (3.2)$$

where  $\partial_i = \partial/\partial q_i$  and  $J^\epsilon$  is the matrix-valued function solving the Lyapunov equation

$$J^\epsilon(q) \Lambda_\epsilon^\star(q) + \Lambda_\epsilon(q) J^\epsilon(q) = \sigma(q) \sigma^\star(q), \quad q \in \mathbb{R}^2.$$

Thanks to (2.4), the equation above has a unique solution  $J^\epsilon$  and it can be explicitly written as

$$\begin{aligned} J^\epsilon(q) &= \int_0^\infty e^{-\Lambda_\epsilon(q)r} \sigma \sigma^\star(q) e^{-\Lambda_\epsilon^\star(q)r} dr \\ &= \int_0^\infty e^{-\lambda(q)Ar} \sigma \sigma^\star(q) e^{\lambda(q)Ar} e^{-2\epsilon r} dr, \quad q \in \mathbb{R}^2. \end{aligned} \quad (3.3)$$

It is immediate to check that

$$e^{-\lambda(q)Ar} = \begin{pmatrix} \cos(\lambda(q)r) & -\sin(\lambda(q)r) \\ \sin(\lambda(q)r) & \cos(\lambda(q)r) \end{pmatrix}, \quad r \geq 0.$$

In what follows, for every  $q \in \mathbb{R}^2$  we denote

$$\begin{pmatrix} a_1(q) & a_0(q) \\ a_0(q) & a_2(q) \end{pmatrix} =: \sigma \sigma^\star(q),$$

and

$$\beta_0(q) := \frac{a_1(q) + a_2(q)}{4}, \quad \beta_1(q) := \frac{a_1(q) - a_2(q)}{4} \quad \beta_2(q) := \frac{a_0(q)}{2}. \quad (3.4)$$

**Lemma 3.1.** *Assume that  $\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}$  is differentiable. Then, there exist  $M : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  and  $R^\epsilon : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  such that for every  $\epsilon > 0$*

$$S^\epsilon(q) = \frac{1}{\epsilon} \frac{\beta_0(q)}{\lambda^2(q)} \nabla^\perp \lambda(q) - M(q) \nabla \lambda(q) + R^\epsilon(q) \nabla \lambda(q), \quad q \in \mathbb{R}^2. \quad (3.5)$$

*Proof.* Thanks to (3.3), we have

$$\begin{aligned} J_{11}^\epsilon(q) &= \frac{\beta_0(q)}{\epsilon} + \beta_1(q) \int_0^\infty \cos(\lambda(q)r) e^{-\epsilon r} dr - \beta_2(q) \int_0^\infty \sin(\lambda(q)r) e^{-\epsilon r} dr \\ J_{22}^\epsilon(q) &= \frac{\beta_0(q)}{\epsilon} - \beta_1(q) \int_0^\infty \cos(\lambda(q)r) e^{-\epsilon r} dr + \beta_2(q) \int_0^\infty \sin(\lambda(q)r) e^{-\epsilon r} dr \\ J_{12}^\epsilon(q) = J_{21}^\epsilon(q) &= \beta_1(q) \int_0^\infty \sin(\lambda(q)r) e^{-\epsilon r} dr + \beta_2(q) \int_0^\infty \cos(\lambda(q)r) e^{-\epsilon r} dr. \end{aligned}$$

Integrating by parts, we have

$$\int_0^\infty \cos(\lambda(q)r) e^{-\epsilon r} dr = \frac{\epsilon}{\lambda^2(q) + \epsilon^2},$$

and

$$\int_0^\infty \sin(\lambda(q)r) e^{-\epsilon r} dr = \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2}.$$

This allows to conclude that

$$\begin{aligned} J_{11}^\epsilon(q) &= \frac{\beta_0(q)}{\epsilon} + \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} - \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \\ J_{22}^\epsilon(q) &= \frac{\beta_0(q)}{\epsilon} - \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \\ J_{12}^\epsilon(q) = J_{21}^\epsilon(q) &= \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2}. \end{aligned} \tag{3.6}$$

Now, due to (3.1), for every  $\epsilon > 0$  and  $q \in \mathbb{R}^2$  we have

$$\begin{aligned} \partial_i (\Lambda_\epsilon^{-1})_{11}(q) &= \partial_i (\Lambda_\epsilon^{-1})_{22}(q) = -\frac{2\epsilon\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \partial_i \lambda(q), \quad i = 1, 2, \\ \partial_i (\Lambda_\epsilon^{-1})_{12}(q) &= -\partial_i (\Lambda_\epsilon^{-1})_{21}(q) = \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} \partial_i \lambda(q), \quad i = 1, 2. \end{aligned} \tag{3.7}$$

Therefore, if we replace (3.6) and (3.7) in (3.2), we obtain

$$\begin{aligned} S_1^\epsilon(q) &= -\frac{2\epsilon\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \left[ \left( \frac{\beta_0(q)}{\epsilon} + \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} - \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \right) \partial_1 \lambda(q) \right. \\ &\quad \left. + \left( \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} \right) \partial_2 \lambda(q) \right] \\ &\quad + \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} \left[ \left( \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} \right) \partial_1 \lambda(q) \right. \\ &\quad \left. + \left( \frac{\beta_0(q)}{\epsilon} - \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \right) \partial_2 \lambda(q) \right], \end{aligned}$$

and

$$\begin{aligned}
S_2^\epsilon(q) = & -\frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} \left[ \left( \frac{\beta_0(q)}{\epsilon} + \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} - \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \right) \partial_1 \lambda(q) \right. \\
& + \left. \left( \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} \right) \partial_2 \lambda(q) \right] \\
& - \frac{2\epsilon \lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \left[ \left( \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} \right) \partial_1 \lambda(q) \right. \\
& \left. + \left( \frac{\beta_0(q)}{\epsilon} - \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} \right) \partial_2 \lambda(q) \right],
\end{aligned}$$

Now, we define

$$\Gamma_1^\epsilon(q) := \beta_1(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2} + \beta_2(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2}, \quad \Gamma_1(q) := \frac{\beta_1(q)}{\lambda(q)},$$

and

$$\Gamma_2^\epsilon(q) := \beta_1(q) \frac{\epsilon}{\lambda^2(q) + \epsilon^2} - \beta_2(q) \frac{\lambda(q)}{\lambda^2(q) + \epsilon^2}, \quad \Gamma_2(q) := -\frac{\beta_2(q)}{\lambda(q)}.$$

With these notations, we have

$$\begin{aligned}
S_1^\epsilon(q) = & \frac{1}{\epsilon} \frac{\beta_0(q)}{\lambda^2(q)} \partial_2 \lambda(q) + \left[ -\frac{2\beta_0(q)}{\lambda^3(q)} + \frac{\Gamma_1(q)}{\lambda^2(q)} \right] \partial_1 \lambda(q) - \frac{\Gamma_2(q)}{\lambda^2(q)} \partial_2 \lambda(q) \\
& + R_{11}^\epsilon(q) \partial_1 \lambda(q) + R_{12}^\epsilon(q) \partial_2 \lambda(q),
\end{aligned}$$

where

$$\begin{aligned}
R_{11}^\epsilon(q) := & -\frac{2\epsilon \lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \Gamma_2^\epsilon(q) + \epsilon \beta_2(q) \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^3} \\
& + 2\beta_0(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \right] - \beta_1(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)(\lambda^2(q) - \epsilon^2)}{(\lambda^2(q) + \epsilon^2)^3} \right], \tag{3.8}
\end{aligned}$$

and

$$\begin{aligned}
R_{12}^\epsilon(q) := & -\frac{2\lambda(q)\epsilon}{(\lambda^2(q) + \epsilon^2)^2} \Gamma_1^\epsilon(q) - \epsilon \beta_2(q) \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^3} \\
& - \frac{\beta_0(q)}{\epsilon} \left[ \frac{1}{\lambda^2(q)} - \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} \right] + \beta_1(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)(\lambda^2(q) - \epsilon^2)}{(\lambda^2(q) + \epsilon^2)^3} \right]. \tag{3.9}
\end{aligned}$$

In a similar way, we have

$$\begin{aligned}
S_2^\epsilon(q) = & -\frac{1}{\epsilon} \frac{\beta_0(q)}{\lambda^2(q)} \partial_1 \lambda(q) - \frac{\Gamma_2(q)}{\lambda^2(q)} \partial_1 \lambda(q) - \left[ \frac{2\beta_0(q)}{\lambda^3(q)} + \frac{\Gamma_1(q)}{\lambda^2(q)} \right] \partial_2 \lambda(q) \\
& + R_{21}^\epsilon(q) \partial_1 \lambda(q) + R_{22}^\epsilon(q) \partial_2 \lambda(q),
\end{aligned}$$

where

$$\begin{aligned}
R_{21}^\epsilon(q) := & -\frac{2\lambda(q)\epsilon}{(\lambda^2(q) + \epsilon^2)^2} \Gamma_1^\epsilon(q) - \epsilon\beta_1(q) \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^3} \\
& + \frac{\beta_0(q)}{\epsilon} \left[ \frac{1}{\lambda^2(q)} - \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} \right] - \beta_2(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)(\lambda^2(q) - \epsilon^2)}{(\lambda^2(q) + \epsilon^2)^3} \right], \tag{3.10}
\end{aligned}$$

and

$$\begin{aligned}
R_{22}^\epsilon(q) := & \frac{2\epsilon\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \Gamma_2^\epsilon - \epsilon\beta_2(q) \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^3} \\
& + 2\beta_0(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} \right] + \beta_1(q) \left[ \frac{1}{\lambda^3(q)} - \frac{\lambda(q)(\lambda^2(q) - \epsilon^2)}{(\lambda^2(q) + \epsilon^2)^3} \right]. \tag{3.11}
\end{aligned}$$

Therefore, recalling that  $\Gamma_1(q) = \beta_1(q)/\lambda(q)$  and  $\Gamma_2(q) = -\beta_2(q)/\lambda(q)$ , if we define

$$M(q) = \frac{1}{\lambda^3(q)} \begin{pmatrix} 2\beta_0(q) - \beta_1(q) & -\beta_2(q) \\ -\beta_2(q) & 2\beta_0(q) + \beta_1(q) \end{pmatrix}, \tag{3.12}$$

and we define  $R^\epsilon(q) = (R_{ij}^\epsilon(q))_{i,j=1,2}$ , where the components  $R_{ij}^\epsilon(q)$  are defined in (3.8), (3.9), (3.10) and (3.11), we obtain (3.5).  $\square$

In what follows we shall assume that the following condition is satisfied.

**Hypothesis 2.** 1. *The mapping  $\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuously differentiable.*

2. *For every  $\epsilon > 0$ , the mapping  $S_\epsilon : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  introduced in (3.2) is locally Lipschitz-continuous and has linear growth.*
3. *For every  $\epsilon > 0$  the mappings  $\Lambda_\epsilon^{-1}b : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  and  $\Lambda_\epsilon^{-1}\sigma : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  are locally Lipschitz-continuous and have linear growth.*

**Remark 3.2.** 1. According to the expression of  $M(q)$  given in (3.12) and the expressions for the coefficients of  $R_\epsilon(q)$  given in (3.8), (3.9), (3.10) and (3.11), thanks to what we have already assumed in Hypothesis 1 we can check easily that Hypothesis 2 is satisfied if we assume  $\sigma$  to be bounded and  $\lambda$  to be bounded and differentiable, with  $\nabla\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  Lipschitz-continuous.

2. In the same way, if we assume that  $\nabla\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is locally Lipschitz-continuous and has linear growth and there exists  $c > 0$  such that for  $|q|$  large enough

$$|\lambda(q)| \geq c|q|^2,$$

then Hypothesis 2 is satisfied, without assuming  $\sigma$  to be bounded.

**Theorem 3.3.** For every  $\mu, \epsilon > 0$ , let  $q_{\mu,\epsilon}$  be the solution of problem (2.7). Then, under Hypotheses 1 and 2, for every  $\epsilon > 0$  we have

$$\lim_{\mu \rightarrow 0} \mathbb{E} \sup_{t \in [0,T]} |q_{\mu,\epsilon}(t) - q_\epsilon(t)| = 0, \quad (3.13)$$

where  $q_\epsilon$  is the solution of the problem

$$dq_\epsilon(t) = [\Lambda_\epsilon^{-1}b(q_\epsilon(t)) + S_\epsilon(q_\epsilon(t))] dt + \Lambda_\epsilon^{-1}\sigma(q_\epsilon(t)) dw(t), \quad q_\epsilon(0) = q. \quad (3.14)$$

*Proof.* According to Hypotheses 1 and 2, we have that for every  $\epsilon > 0$  and for every  $k \geq 1$  and  $T > 0$  problem (3.14) admits a unique solution  $q_\epsilon \in L^k(\Omega; C([0,T]; \mathbb{R}^2))$ . As  $\langle \Lambda_\epsilon(q)p, p \rangle = \epsilon |p|^2$ , this allows to conclude thanks to [12, Theorem 2.4].  $\square$

## 4 The averaging limit

In this section we want to investigate the limiting behavior of the slow component of  $q_\epsilon$ , as  $\epsilon$  goes to zero. To this purpose, we need to introduce some preliminary material.

### 4.1 Some notations and further assumptions

We consider here the system

$$\dot{X}(t) = \frac{\beta_0(X(t))}{\lambda^2(X(t))} \nabla^\perp \lambda(X(t)). \quad (4.1)$$

Clearly, for every  $t \geq 0$ , we have  $\lambda(X(t)) = \lambda(X(0))$ . Now, if we consider the perturbed system

$$\begin{aligned} dX_\epsilon(q) &= \frac{\beta_0(X_\epsilon(t))}{\lambda^2(X_\epsilon(t))} \nabla^\perp \lambda(X_\epsilon(t)) dt \\ &+ \epsilon \left[ \frac{1}{\lambda(X_\epsilon(t))} Ab(X_\epsilon(t)) - M(X_\epsilon(t)) \nabla \lambda(X_\epsilon(t)) \right] dt + \frac{\sqrt{\epsilon}}{\lambda(X_\epsilon(t))} A\sigma(X_\epsilon(t)) dw(t) \\ &+ \epsilon^2 \left[ H^\epsilon(X_\epsilon(t))b(X_\epsilon(t)) + \hat{R}^\epsilon(X_\epsilon(t)) \nabla \lambda(X_\epsilon(t)) \right] dt + \epsilon H^\epsilon(X_\epsilon(t))\sigma(X_\epsilon(t)) dw(t), \end{aligned}$$

the quantity  $\lambda(X_\epsilon(t))$  is not anymore conserved. However, for any fixed time interval  $[0, T]$  and for every  $k \geq 1$ , we have

$$\lim_{\epsilon \rightarrow 0} \mathbb{E} \sup_{t \in [0,T]} |X_\epsilon(t) - X(t)|^k = 0,$$

and, as an immediate consequence,

$$\lim_{\epsilon \rightarrow 0} \mathbb{E} \sup_{t \in [0,T]} |\lambda(X_\epsilon(t)) - \lambda(X(0))|^k = 0.$$

Now, with the change of time  $t \mapsto t/\epsilon$ , we can check that

$$\mathcal{L}(X_\epsilon(\cdot/\epsilon)) = \mathcal{L}(q_\epsilon(\cdot)),$$

where  $q_\epsilon$  is the solution of equation (3.14). As we mentioned above, our aim is to identify the non trivial limit for the distribution of the process  $\lambda(q_\epsilon(\cdot))$ , as  $\epsilon \downarrow 0$ . To this purpose, in addition to Hypotheses 1 and 2, we assume that  $\lambda$  satisfies the following conditions.

**Hypothesis 3.** 1. If  $\beta_0$  is the function defined in (3.4), we have

$$\inf_{x \in \mathbb{R}^2} \beta_0(x) > 0. \quad (4.2)$$

2. The mapping  $\lambda : \mathbb{R}^2 \rightarrow \mathbb{R}$  is four times continuously differentiable, with bounded second derivative.
3. The mapping  $\lambda$  has only a finite number of critical points  $x_1, \dots, x_n$ . The matrix of second derivatives  $D^2\lambda(x_i)$  is non degenerate, for every  $i = 1, \dots, n$  and  $\lambda(x_i) \neq \lambda(x_j)$ , if  $i \neq j$ .
4. There exist three positive constants  $a_1, a_2, a_3$  such that  $\lambda(x) \geq a_1 |x|^2$ ,  $|\nabla \lambda(x)| \geq a_2 |x|$  and  $\Delta \lambda(x) \geq a_3$ , for all  $x \in \mathbb{R}^2$ , with  $|x|$  large enough.

**Remark 4.1.** Remember that the function  $\beta_0$  was defined as  $[(\sigma\sigma^*)_{11}^2 + (\sigma\sigma^*)_{22}^2]/4$ . Therefore, condition (4.2) is a non-degeneracy condition on the noisy perturbation.

Next, for every  $z \geq \lambda_0$ , we denote by  $C(z)$  the  $z$ -level set

$$C(z) = \{x \in \mathbb{R}^2 : \lambda(x) = z\}.$$

The set  $C(z)$  may consist of several connected components

$$C(z) = \bigcup_{k=1}^{N(z)} C_k(z),$$

and for every  $x \in \mathbb{R}^2$  we have

$$X(0) = x \implies X(t) \in C_{k(x)}(\lambda(x)), \quad t \geq 0,$$

where  $C_{k(x)}(x)$  is the connected component of the level set  $C(\lambda(x))$ , to which the point  $x$  belongs. For every  $z \geq 0$  and  $k = 1, \dots, N(z)$ , we shall denote by  $G_k(z)$  the domain of  $\mathbb{R}^2$  bounded by the level set component  $C_k(z)$ .

If we identify all points in  $\mathbb{R}^2$  belonging to the same connected component of a given level set  $C(z)$  of the Hamiltonian  $\lambda$ , we obtain a graph  $\Gamma$ , given by several intervals  $I_1, \dots, I_n$  and vertices  $O_1, \dots, O_m$ . The vertices will be of two different types, external and internal vertices. External vertices correspond to local extrema of  $\lambda$ , while internal vertices correspond to saddle points of  $\lambda$ . Among external vertices, we will also include  $O_\infty$ , the endpoint of the interval in the graph corresponding to the point at infinity.

In what follows, we shall denote by  $\Pi : \mathbb{R}^2 \rightarrow \Gamma$  the *identification map*, that associates to every point  $x \in \mathbb{R}^2$  the corresponding point  $\Pi(x)$  on the graph  $\Gamma$ . We have  $\Pi(x) = (\lambda(x), k(x))$ , where  $k(x)$  denotes the number of the interval on the graph  $\Gamma$ , containing the point  $\Pi(x)$ . If  $O_i$  is one of the interior vertices, the second coordinate cannot be chosen in a unique way, as there are three edges having  $O_i$  as their endpoint. Notice that both  $k(x)$  and  $H(x)$  are first integrals (a discrete and a continuous one, respectively) for system (4.1).

On the graph  $\Gamma$ , a distance can be introduced in the following way. If  $y_1 = (z_1, k)$  and  $y_2 = (z_2, k)$  belong to the same edge  $I_k$ , then  $d(y_1, y_2) = |z_1 - z_2|$ . In the case  $y_1$  and  $y_2$  belong to different edges, then

$$d(y_1, y_2) = \min \left\{ d(y_1, O_{i_1}) + d(O_{i_1}, O_{i_2}) + \cdots + d(O_{i_j}, y_2) \right\},$$

where the minimum is taken over all possible paths from  $y_1$  to  $y_2$ , through every possible sequence of vertices  $O_{i_1}, \dots, O_{i_j}$ , connecting  $y_1$  to  $y_2$ .

If  $z$  is not a critical value, then each  $C_k(z)$  consists of one periodic trajectory of the vector field  $\nabla^\perp \lambda(x)$ . If  $z$  is a local extremum of  $\lambda(x)$ , then, among the components of  $C(z)$  there is a set consisting of one point, the rest point of the flow. If  $\lambda(x)$  has a saddle point at some point  $x_0$  and  $\lambda(x_0) = z$ , then  $C(z)$  consists of three trajectories, the equilibrium point  $x_0$  and the two trajectories that have  $x_0$  as their limiting point, as  $t \rightarrow \pm\infty$ .

Now, for every  $(z, k) \in \Gamma$ , we define

$$T_k(z) = \oint_{C_k(z)} \frac{\lambda^2(x)}{\beta_0(x)|\nabla\lambda(x)|} dl_{z,k}, \quad (4.3)$$

where  $dl_{z,k}$  is the length element on  $C_k(z)$ . Notice that  $T_k(z)$  is the period of the motion along the level set  $C_k(z)$ .

As we have seen above, if  $X(0) = x \in C_k(z)$ , then  $X(t) \in C_k(z)$ , for every  $t \geq 0$ . As known, for every  $(z, k) \in \Gamma$  the probability measure

$$d\mu_{z,k} := \frac{1}{T_k(z)} \frac{\lambda^2(x)}{\beta_0(x)|\nabla\lambda(x)|} dl_{z,k} \quad (4.4)$$

is invariant for system (4.1) on the level set  $C_k(z)$ .

## 4.2 The limit of $\Pi(q_\epsilon)$

Due to (3.1), for every  $\epsilon > 0$  we have

$$\Lambda_\epsilon^{-1}(q) = \frac{1}{\lambda(q)} A + \epsilon H^\epsilon(q), \quad (4.5)$$

where

$$H^\epsilon(q) := \frac{1}{\lambda^2(q) + \epsilon^2} \left( I - \frac{\epsilon}{\lambda(q)} A \right).$$

Notice that

$$\sup_{\epsilon > 0} |H^\epsilon(q)| < \infty, \quad q \in \mathbb{R}^2. \quad (4.6)$$

**Lemma 4.2.** *Let  $R^\epsilon : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$  be the mapping introduced in Lemma 3.1. Then*

$$\sup_{\epsilon > 0} \frac{1}{\epsilon} |\hat{R}^\epsilon(q)| < \infty, \quad q \in \mathbb{R}^2. \quad (4.7)$$

*Proof.* We have

$$\frac{1}{\lambda^3(q)} - \frac{\lambda(q)}{(\lambda^2(q) + \epsilon^2)^2} = \epsilon^2 \left[ \frac{2\lambda^2(q) + \epsilon^2}{\lambda^3(q)(\lambda^2(q) + \epsilon^2)^2} \right],$$

and

$$\frac{1}{\lambda^3(q)} - \frac{\lambda(q)(\lambda^2(q) - \epsilon^2)}{(\lambda^2(q) + \epsilon^2)^3} = \epsilon^2 \left[ \frac{4\lambda^4(q) + 3\epsilon^2\lambda^2(q) + \epsilon^4}{\lambda^3(q)(\lambda^2(q) + \epsilon^2)^3} \right]$$

and

$$\frac{1}{\lambda^2(q)} - \frac{\lambda^2(q) - \epsilon^2}{(\lambda^2(q) + \epsilon^2)^2} = \epsilon^2 \left[ \frac{3\lambda^2(q) + \epsilon^2}{\lambda^2(q)(\lambda^2(q) + \epsilon^2)^2} \right].$$

Therefore, recalling how  $R^\epsilon(q)$  was defined in (3.8), (3.9), (3.10) and (3.11), we can conclude.  $\square$

According to (3.5), (4.5), (4.6) and (4.7), equation (3.14) can be rewritten as

$$dq_\epsilon(q) = \frac{1}{\epsilon} \frac{\beta_0(q_\epsilon(t))}{\lambda^2(q_\epsilon(t))} \nabla^\perp \lambda(q_\epsilon(t)) dt + B(q_\epsilon(t)) dt + \Sigma(q_\epsilon(t)) dw(t), \quad (4.8)$$

$$+ \epsilon [B_\epsilon(q_\epsilon(t)) dt + \Sigma_\epsilon(q_\epsilon(t)) dw(t)], \quad q_\epsilon(0) = q,$$

where

$$B(q) = \frac{1}{\lambda(q)} Ab(q) - M(q) \nabla \lambda(q), \quad \Sigma(q) = \frac{1}{\lambda(q)} A\sigma(q),$$

and

$$B_\epsilon(q) = H^\epsilon(q)b(q) + \frac{1}{\epsilon} R^\epsilon(q) \nabla \lambda(q), \quad \Sigma_\epsilon(q) = H^\epsilon(q)\sigma(q).$$

This means that, as  $\epsilon \downarrow 0$ , part of the coefficients are of order  $O(\epsilon^{-1})$ , part of order  $O(1)$  and part of order  $O(\epsilon)$ .

With the notations introduced in the previous section, in what follows, we want to investigate the limiting behavior of the  $\Gamma$ -valued process  $\Pi(q_\epsilon(\cdot)) = (\lambda(q_\epsilon(\cdot)), k(q_\epsilon(\cdot)))$ , as  $\epsilon \downarrow 0$ .

If we apply Itô's formula to  $\lambda(q_\epsilon(t))$ , we get

$$d\lambda(q_\epsilon(t)) = \mathcal{G}\lambda(q_\epsilon(t)) dt + \mathcal{A}\lambda(q_\epsilon(t)) dw(t) + \epsilon \mathcal{G}_\epsilon\lambda(q_\epsilon(t)) dt + \epsilon \mathcal{A}_\epsilon\lambda(q_\epsilon(t)) dw(t),$$

where for every  $f \in C^2(\mathbb{R}^2)$  and  $q \in \mathbb{R}^2$

$$\mathcal{G}f(q) = \frac{1}{2} \text{Tr} [\Sigma \Sigma^\star(q) D^2 f(q)] + \langle Df(q), B(q) \rangle,$$

$$\mathcal{A}f(q) = \Sigma(q)^\star Df(q),$$

$$\mathcal{G}_\epsilon f(q) = \frac{1}{2} \text{Tr} [(\epsilon \Sigma_\epsilon \Sigma_\epsilon^\star(q) + \Sigma \Sigma_\epsilon^\star(q) + \Sigma_\epsilon \Sigma^\star(q)) D^2 f(q)] + \langle Df(q), B_\epsilon(q) \rangle,$$

and

$$\mathcal{A}_\epsilon f(q) = \Sigma_\epsilon^\star(q) Df(q).$$

We recall that the graph  $\Gamma$  is made of  $n$  intervals  $I_1, \dots, I_n$  and  $m$  vertices  $O_1, \dots, O_m$ . For every  $j = 1, \dots, n$  and for every  $f$  that is twice differentiable in the interior of the edge  $I_j$ , we denote

$$\mathcal{L}_j f(z) = \frac{1}{2} \alpha_j(z) f''(z) + \gamma_j(z) f'(z), \quad (4.9)$$

where

$$\begin{aligned}\alpha_j(z) &= \oint_{C_j(z)} |\mathcal{A}\lambda(x)|^2 d\mu_{z,j}(x) = \oint_{C_j(z)} |\Sigma^*(x)\nabla\lambda(x)|^2 d\mu_{z,j}(x), \\ \gamma_j(z) &= \oint_{C_j(z)} \mathcal{G}\lambda(x) d\mu_{z,j}(x),\end{aligned}$$

and  $d\mu_{z,j}$  is the probability measure introduced in (4.4).

**Definition 4.3.** For each interior vertex  $O_k$  and any segment  $I_j$  meeting at  $O_k$  (notation  $I_j \sim O_k$ ), let  $\rho_{kj}$  be the positive constant defined by

$$\rho_{kj} = \oint_{C_{kj}} \frac{\lambda^2(x)}{\beta_0(x)|\nabla\lambda(x)|} |\Sigma^*(x)\nabla\lambda(x)|^2 dl(x).$$

We denote by  $D(L) \subset C(\Gamma)$  the set consisting all continuous functions  $f$  defined on the graph  $\Gamma$  such that  $\mathcal{L}_j f$  is well defined in the interior of the edge  $I_j$  and for every  $I_j \sim O_k$  there exists finite

$$\lim_{x \rightarrow O_k} \mathcal{L}_j f(x)$$

and the limit is independent of the edge  $I_j$ . Moreover, for each interior vertex  $O_k$

$$\sum_{j: I_j \sim O_k} \pm \rho_{kj} f'_j(\lambda(O_k)) = 0,$$

where  $f'_j$  denotes the derivative of  $f$  with respect to the local coordinate  $\lambda$ , along the edge  $I_j$  and the sign  $\pm$  are taken if  $\lambda > \lambda(O_k)$  or  $\lambda < \lambda(O_k)$ .

Next, for every  $f \in D(L)$ , we define

$$Lf(x) = \begin{cases} \mathcal{L}_j f(x), & \text{if } x \text{ is an interior point of } I_j, \\ \lim_{x \rightarrow O_k} \mathcal{L}_j f(x), & \text{if } x \text{ is the vertex } O_k \text{ and } I_j \sim O_k. \end{cases}$$

As proven in [11, Theorem 8.2.1], in case  $\Sigma(q) = I$  the operator  $L$  defined on the domain  $D(L)$ , as described in Definition 4.3, is the generator of a strong Markov process  $Y_t$  on  $\Gamma$  with continuous trajectories. Here the same result holds, because of the non-degeneracy condition (4.2) satisfied by the diffusion coefficient  $\Sigma(q)$ .

In fact, as shown in the next theorem, the MArkov process  $Y$  is the weak limit in  $C([0, T]; \Gamma)$  of the slow motion  $\Pi(q_\epsilon(\cdot))$  on  $\Gamma$ .

**Theorem 4.4.** Under Hypotheses 1, 2 and 3, for every fixed  $T > 0$  the  $\Gamma$ -valued process  $\Pi(q_\epsilon(\cdot))$  converges weakly in  $C([0, T]; \Gamma)$  to the Markov process  $Y$  generated by the operator  $(L, D(L))$ , introduced in Definition 4.3.

*Proof.* In case in equation (4.8) we have  $B(q) = B_\epsilon(q) = \Sigma_\epsilon(q) = 0$  and  $\Sigma(q) = I$ . the result above is what is proven in [11, Theorem 8.2.2]. In the present situation we are dealing with the more general situation in which we have a coefficient  $B(q)$  of order  $O(1)$  and coefficients  $B_\epsilon(q)$  of order  $O(\epsilon)$ . Moreover we allow a non-constant diffusion coefficient  $\Sigma(q) + \Sigma_\epsilon(q)$ , where  $\Sigma(q)$  is of order  $O(1)$  and  $\Sigma_\epsilon(q)$  is of order  $O(\epsilon)$ . As shown in [24], under these more general assumptions, an averaging principle of the same type of the one described in [11, Theorem

8.2.2] is still valid. This of course has required to introduce a suitable generalization of the operator  $(L, D(L))$ , that takes into account the coefficients  $B$  and  $\Sigma$ , and to extend the limiting result in presence of the vanishing terms  $B_\epsilon$  and  $\Sigma_\epsilon$ .

□

## References

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