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# Scaling limit of a kinetic inhomogeneous stochastic system in the quadratic potential

Thomas Cavallazzi and Emeline Luirard

Univ Rennes, CNRS, IRMAR - UMR 6625, F-35000 Rennes, France {Thomas.Cavallazzi,Emeline.Luirard}@univ-rennes1.fr

Abstract: We consider a particle evolving in the quadratic potential and subject to a time-inhomogeneous frictional force and to a random force. The couple of its velocity and position is solution to a stochastic differential equation driven by an  $\alpha$ -stable Lévy process with  $\alpha \in (1,2]$  and the frictional force is of the form  $t^{-\beta} \operatorname{sgn}(v)|v|^{\gamma}$ . We identify three regimes for the behavior in long-time of the couple velocity-position with a suitable rescaling, depending on the balance between the frictional force and the index of stability  $\alpha$  of the noise.

**Keywords:** kinetic stochastic equation; damping Hamiltonian system; time-inhomogeneous stochastic differential equation; Lévy process; scaling transformation; asymptotic distribution; asymptotic expansion of solutions to ordinary differential equations.

MSC2020 Subject Classification: Primary 60H10; Secondary 60F17; 60G52; 60G18; 60J65; 34E05.

## 1 Introduction and main results

In this paper, we study the long-time behavior of a stochastic system modelling a particle, with velocity  $V \in \mathbb{R}$  and position  $X \in \mathbb{R}$ . The particle evolves in the quadratic potential  $\mathcal{U}: x \mapsto \frac{x^2}{2}$ , and is subject to a time-inhomogeneous frictional force b and to a random force, representing the interaction with the surrounding heat bath. The dynamics of the particle is described by the stochastic damping Hamiltonian system, driven by an  $\alpha$ -stable process L with  $\alpha \in (1, 2]$ ,

$$\begin{cases}
dV_t = dL_t - b(t, V_t) dt - \nabla \mathcal{U}(X_t) dt, \\
dX_t = V_t dt, \\
(V_{t_0}, X_{t_0}) = (v_0, x_0).
\end{cases}$$
(1)

Stochastic Hamiltonian systems have been widely studied in the time-homogeneous setting, i.e. taking b=0 in (1). An interesting problem is to understand their asymptotic behaviors. The Hamiltonian process associated with this system is defined by  $H_t := \frac{1}{2}|V_t|^2 + \mathcal{U}(X_t)$ . For example, the long-time behavior of the Hamiltonian process under a suitable rescaling has been studied in [AK94]. The case of time-homogeneous damping Hamiltonian systems is tackled in [Wu01] (see also references therein).

The long-time behavior of a particle evolving in a free potential, i.e.  $\mathcal{U}=0$ , has already been studied, see e.g. [GO13], [FT21], [GL21a], [GL21b] and references therein.

Even in the time-homogeneous case, various asymptotic behaviors can appear. Whenever the

random force is supposed to be Brownian, i.e.  $\alpha = 2$ , a particular non-linear Langevin's type SDE was studied in [FT21]:

$$V_t = v_0 + B_t - \frac{\rho}{2} \int_0^t \frac{V_s}{1 + V_s^2} ds$$
 and  $X_t = x_0 + \int_0^t V_s ds$ .

Using a suitable rescaling, the authors show that the position process behaves asymptotically as a Brownian motion, a stable process, or an integrated Bessel process, depending on the value of  $\rho$ . For example, in the normal diffusive regime, the rescaled velocity-position process behaves asymptotically as an ergodic process and a Brownian motion. In the integrated Bessel regime, the Bessel process and its time integral are the asymptotic distributions. However, the tools used therein, such as invariant measure, scale function and speed measure, are limited to time-homogeneous coefficients.

In [GO13] and [GL21a], the drift coefficient b was allowed to depend on time under a homogeneity condition. More precisely, the following system is considered

$$\begin{cases} dV_t = dB_t - \rho \frac{\operatorname{sgn}(V_t) |V_t|^{\gamma}}{t^{\beta}} dt, \\ dX_t = V_t dt. \end{cases}$$

In [GO13], the authors study the convergence in distribution, when t tends to  $+\infty$ , of  $r_t V_t$ , for a certain rate of convergence  $r_t$ . In [GL21a], the authors extend the results obtained in [GO13] to the whole process given by the couple velocity-position. Namely, the authors study the limit in distribution of the rescaled process  $(r_{\varepsilon,V}V_{t/\varepsilon}, r_{\varepsilon,X}X_{t/\varepsilon})_t$  for two appropriate rates of convergence  $r_{\varepsilon,V}$  and  $r_{\varepsilon,X}$ . Results were further generalized in [GL21b] to an  $\alpha$ -stable driving process. To be more precise, the authors highlight three regimes, depending on the balance between  $\beta$ ,  $\gamma$  and  $\alpha$ , the index of stability of L. Whenever the frictional force is sufficiently "small at infinity",i.e. if  $\beta$  is large enough, the rescaled process behaves as if there was no frictional force and thus converges in distribution towards the Kolmogorov process  $(\mathcal{L}, \int_0^{\cdot} \mathcal{L})$ , where  $\mathcal{L}$  has the same distribution as the noise. When the two forces offset, the limiting process is still of kinetic form  $(\mathcal{V}, \int_0^{\cdot} \mathcal{V})$ , but the process  $\mathcal{V}$  is henceforth ergodic. Whereas, when the drag force swings with the random process, the limiting process is no longer kinetic. The rescaled velocity process converges in finite dimensional distributions towards a white noise. The proofs are essentially based on the self-similarity of the driving process and on moment estimates of the velocity process.

Degenerate systems has been intensively studied for several years. In particular, the existence and uniqueness of solutions to degenerate SDEs have been discussed in many papers. These models are called degenerate because the noise is only present in one component of the system but can be transferred into others by drift terms. The well-posedness of these systems, when their deterministic version is ill-posed, can be proved by taking advantage of the regularizing effect of the noise and of its propagation through the whole system. The case of Brownian degenerate SDEs has been of course wildly explored, see e.g. [FFPV17], [WZ15], [Zha16], [dR17], [HMC18] and references therein. The time-dependence is treated in the last four cited papers.

The case of a Lévy driving process is more recent, see e.g. [Zha14] for a time-homogeneous setting, and [MM21] for drifts depending on time.

In this paper, we are interested in the long-time behavior of the solution to the following

system of SDEs, defined on the time interval  $[t_0, +\infty)$ , where  $t_0 > 0$ 

$$\begin{cases}
dV_t = dL_t - \operatorname{sgn}(V_t) \frac{|V_t|^{\gamma}}{t^{\beta}} dt - X_t dt, \\
dX_t = V_t dt, \\
(V_{t_0}, X_{t_0}) = (v_0, x_0).
\end{cases}$$
(SKE)

The frictional force is time-inhomogeneous, depending on non-negative parameters  $\beta$  and  $\gamma$ . The driving process L is either a Brownian motion, i.e.  $\alpha=2$ , or a symmetric  $\alpha$ -stable Lévy process with  $\alpha\in(1,2)$ . More precisely, our goal is to study the asymptotic behavior, as  $\varepsilon\to0$ , of the rescaled velocity-position process  $(Z_t^{(\varepsilon)})_t:=\left(r_\varepsilon(X_{t/\varepsilon},V_{t/\varepsilon})^T\right)_t$ , for an appropriate rate of convergence  $r_\varepsilon$ . One of our motivations is to study how the presence of the quadratic potential influences the results obtained in [GL21a, GL21b] through a confining effect on the position X. Our system without noise and frictional force is nothing else than the classical harmonic oscillator

$$\begin{cases} v_t' = -x_t, \\ x_t' = v_t. \end{cases}$$

The intrinsic oscillatory behavior induced by the quadratic potential prevents the rescaled process  $Z^{(\varepsilon)}$  from converging as a process. However, we prove that each of its one-dimensional marginal distributions converges. In order to obtain the convergence of the whole process, the key idea is to remove the oscillations present in the system. Namely, we set  $Y_t := \Theta_t^{-1}(X_t, V_t)^T$ , where  $\Theta_t$  is the rotation on  $\mathbb{R}^2$  of angle t defined by

$$\Theta_t := \begin{pmatrix} \cos(t) & -\sin(t) \\ \sin(t) & \cos(t) \end{pmatrix}.$$

Let us now state the main results of our paper. The following theorem deals with convergences in distribution in the space of continuous functions  $\mathcal{C}((0,+\infty),\mathbb{R})$ .

**Theorem 1.1** (Brownian case, i.e.  $\alpha = 2$ ). Define  $q := \frac{\beta}{\gamma + 1}$ ,  $r_{\varepsilon} := \varepsilon^{q \wedge \frac{1}{2}}$  and set  $(Y_t^{(\varepsilon)})_{t \geq \varepsilon t_0} := \left(r_{\varepsilon}\Theta_{t/\varepsilon}^{-1}(X_{t/\varepsilon}, V_{t/\varepsilon})^T\right)_{t \geq \varepsilon t_0}$ . Let  $\mathcal{B}$  be a standard two-dimensional Brownian motion on  $\mathbb{R}^2$ .

- (i) (Super-critical regime i.e. 2q > 1). The rescaled process  $Y^{(\varepsilon)}$  converges in distribution towards  $\left(\mathcal{B}_{\frac{t}{2}}\right)_{t>0}$ .
- (ii) (Critical regime i.e. 2q = 1). Assume that  $\gamma = 1$ . The rescaled process  $Y^{(\varepsilon)}$  converges in distribution towards  $\left(\frac{1}{\sqrt{2t}}\int_0^t \sqrt{s}\,\mathrm{d}\mathcal{B}_s\right)_{t>0}$ , which is the centered Gaussian process with covariance kernel  $K(s,t) = \frac{(s\wedge t)^2}{4\sqrt{st}}I_2$ .
- (iii) (Sub-critical regime i.e. 2q < 1). Assume that  $\gamma = 1$  and  $\beta > \frac{1}{2}$ . The rescaled process  $Y^{(\varepsilon)}$  converges in finite dimensional distributions towards the centered Gaussian process with covariance kernel  $K(s,t) = \frac{1}{2} s^{\beta} \mathbb{1}_{\{s=t\}} I_2$ .

Let us denote by  $\psi$  the characteristic exponent of the symmetric stable process L. It follows from Theorem 14.15 p. 86 in [Sat99] that there exists a > 0 such that for all  $\xi \in \mathbb{R}$ ,

$$\psi(\xi) = -a|\xi|^{\alpha}.\tag{2}$$

In the next theorem, the convergences occur in the space of right-continuous with left limits (càdlàg) functions  $\mathcal{D}((0,+\infty),\mathbb{R})$ , i.e. the functions defined on  $(0,+\infty)$  which are càdlàg on every compact subinterval of  $(0,+\infty)$ , which is endowed with the Skorokhod metric.

**Theorem 1.2** (Stable case, i.e.  $\alpha \in (1,2)$ ). Assume that  $\gamma \in (0,\alpha)$ . Define  $q := \frac{\beta}{\gamma + \alpha - 1}$ ,  $r_{\varepsilon} := \varepsilon^{q \wedge \frac{1}{\alpha}}$  and set  $(Y_t^{(\varepsilon)})_{t \geq \varepsilon t_0} := \left(r_{\varepsilon} \Theta_{t/\varepsilon}^{-1} (X_{t/\varepsilon}, V_{t/\varepsilon})^T\right)_{t \geq \varepsilon t_0}$ . Let  $\mathcal{L}$  be a rotationally invariant stable process on  $\mathbb{R}^2$ , whose characteristic exponent is given by

$$\xi \in \mathbb{R}^2 \mapsto -\widetilde{C} \|\xi\|^{\alpha}, \quad \text{with } \widetilde{C} := \frac{a}{2\pi} \int_0^{2\pi} |\cos(x)|^{\alpha} dx.$$

- (i) (Super-critical regime i.e.  $\alpha q > 1$ ). The rescaled process  $Y^{(\varepsilon)}$  converges in distribution towards  $(\mathcal{L}_t)_{t>0}$ .
- (ii) (Critical regime i.e.  $\alpha q = 1$ ). Assume that  $\gamma = 1$ . The rescaled process  $Y^{(\varepsilon)}$  converges in distribution towards the Lévy-type process  $\left(\frac{1}{\sqrt{t}}\int_0^t \sqrt{s} \,\mathrm{d}\mathcal{L}_s\right)_{t>0}$ .
- (iii) (Sub-critical regime i.e.  $\alpha q < 1$ ). Assume that  $\gamma = 1$  and  $\beta > \frac{1}{2}$ . Then, for all  $(t_1, \dots, t_d) \in (0, +\infty)^d$ ,  $(Y_{t_1}^{(\varepsilon)}, \dots, Y_{t_d}^{(\varepsilon)})$  converges in distribution towards the product measure  $\mu_{t_1} \otimes \dots \otimes \mu_{t_d}$ , where  $\mu_t$  is the distribution with characteristic function

$$\xi \in \mathbb{R}^2 \mapsto \exp\left(-\frac{2}{\alpha}\widetilde{C} \|\xi\|^{\alpha} t^{\beta}\right).$$

**Remark 1.3.** The symmetry of  $\nu$  is only required to ensure the well-posedness of (SKE) when  $\gamma < 1$ . Furthermore, the results of the preceding theorem are still true for the finite-dimensional convergence for an  $\alpha$ -stable Lévy driving process with  $\alpha \in (0,1]$ .

Remark 1.4. The Hamiltonian process associated with the system is given by

$$H_t := \frac{1}{2} |V_t|^2 + \frac{1}{2} |X_t|^2 = \frac{1}{2} ||Z_t||^2 = \frac{1}{2} ||Y_t||^2.$$

Combining the preceding results with the continuous mapping theorem, we deduce the convergence of the rescaled energy process  $(H_t^{(\varepsilon)})_{t>0} := (r_\varepsilon^2 H_{t/\varepsilon})_{t>0}$  as  $\varepsilon \to 0$  either as a process in the critical and super-critical regimes, or for finite dimensional distributions in the sub-critical regime. For example in the super-critical regime with  $\alpha=2$ , the limiting energy process  $(H_t^0)_{t\geq 0} := (\frac{1}{2}\|\mathcal{B}_{\frac{t}{2}}\|^2)_{t\geq 0}$  is the squared Bessel process, which is the solution to the following equation

$$dH_t^0 = \sqrt{H_t^0} dB_t + \frac{1}{2} dt, \quad H_0^0 = 0,$$

where B is a standard one-dimensional Brownian motion. Note that we recover the limiting energy process obtained in [AK94](see Theorem 2.1) for the non-damped Hamiltonian system. The explanation is that if the frictional force decreases sufficiently quickly as  $t \to +\infty$ , namely if  $\beta$  is large enough, then the rescaled Hamiltonian process converges as if there were no damping.

We obtain furthermore the convergence in distribution as  $t \to +\infty$  of  $t^{-q \wedge \frac{1}{\alpha}} (X_t, V_t)^T$  in the following corollary.

Corollary 1.5. Let us define  $(Z_t^{(\varepsilon)})_{t \geq \varepsilon t_0} := (r_{\varepsilon}(X_{t/\varepsilon}, V_{t/\varepsilon})^T)_{t \geq \varepsilon t_0}$ , where  $r_{\varepsilon} := \varepsilon^{q \wedge \frac{1}{\alpha}}$ . The rescaled process  $Z^{(\varepsilon)}$  does not converge in distribution. However, we deduce from Theorems 1.1 and 1.2 and under the same assumptions, the convergence in distribution of  $r_{1/t}(X_t, V_t)^T$  towards explicit limits, as  $t \to +\infty$ .

In the Brownian case, the limit is either  $\mathcal{N}(0, \frac{1}{2}I_2)$  in the super-critical and sub-critical regimes, or  $\mathcal{N}(0, \frac{1}{4}I_2)$  in the critical regime.

In the stable case, keeping the same notations as in the preceding theorems, the characteristic function of the limit is given, for all  $\xi \in \mathbb{R}^2$ , by

(i) 
$$\exp\left(-\widetilde{C}\left\|\xi\right\|^{\alpha}\right)$$
 in the super-critical regime,

(ii) 
$$\exp\left(-\left(1+\frac{\alpha}{2}\right)^{-1}\widetilde{C}\left\|\xi\right\|^{\alpha}\right)$$
 in the critical regime,

(iii) 
$$\exp\left(-\frac{2}{\alpha}\widetilde{C}\left\|\xi\right\|^{\alpha}\right)$$
 in the sub-critical regime.

The switch between the three regimes results in different scale parameters of the limiting distributions.

Let us also notice that in the Brownian setting, the position X and the velocity V become independent in large time since the covariance matrix of the limiting Gaussian distribution is diagonal. However, this is false in the stable case. Indeed, the limit is a rotationally invariant stable distribution on  $\mathbb{R}^2$ , which cannot have independent coordinates.

As in [GL21a, GL21b], we highlight three regimes. However, the rate of convergence of the position X is different from that found in [GL21a, GL21b], when  $\mathcal{U}=0$ . Indeed, contrary to the free potential system, the position process is somehow more diffusive. This is due to the structure of our model. Namely, the presence of the quadratic potential allows the noise to propagate more efficiently from the velocity component to the position one (see [FFPV17] for more details). This explains why both the limiting processes and the rate of convergence are different between our work and [GL21a, GL21b]. Let us also note that the position process grows more slowly in our case than when  $\mathcal{U}=0$ . For example, in the Brownian super-critical regime,  $X_t$  behaves asymptotically as  $\mathcal{N}(0,\frac{t}{2})$  in our framework, but as  $\mathcal{N}(0,\frac{t^3}{3})$  in the free potential one. This difference can also be seen in moment estimates established for the position process X (see Remarks 2.3 and 3.4). This is explained by the fact that the quadratic potential confines the particle through a spring force.

In our model, the particle is no longer free. Consequently, both equations are intrinsically linked to each other. Therefore, we can no longer separate by components the study of the velocity-position process. We thus write the system (SKE) in a vector viewpoint, as done in [FFPV17], and use a variation of constants method to return to the study of a two-dimensional system in a free potential. We then adapt the methods used in [GL21a, GL21b]. In the supercritical regime, the proof essentially relies of the self-similarity of the driving process and on moment estimates of V and X. In the critical and sub-critical regimes, we need to restrict ourselves to a linear drag force, i.e.  $\gamma=1$ , in order to rely on the study of the asymptotic behavior of the underlying non-autonomous ordinary differential equation (ODE). Whenever the driving process is Brownian, we take advantage of the theory of Gaussian processes. The convergence is thus characterized by the study of the mean and covariance functions. In the case of a stable driving process, we need to study the convergence, in distribution and as a process, of a Wiener-Lévy integral (i.e. the integral of a deterministic function integrated against a stable process). The key point here is to use the fact that a Wiener-Lévy integral is a process with independent increments.

Our paper is organized as follows. We consider the case of a Brownian driving process in Section 2, and we follow the same structure for an  $\alpha$ -stable driving process in Section 3. For the

sake of clarity, we opt for separating the two cases since the tools used are different. Finally, we state and prove some technical results in Appendix A and Appendix B.

For simplicity, we shall write  $\mathcal{C}$  and  $\mathcal{D}$  for  $\mathcal{C}((0,+\infty),\mathbb{R})$  and  $\mathcal{D}((0,+\infty),\mathbb{R})$ , respectively. For  $x,y\in\mathbb{R}^2$ ,  $\|x\|$  represent the Euclidean norm of x, and  $x\cdot y$  the inner product of x and y. If  $x\in\mathbb{R}^2$ , for each  $i\in\{1,2\}$ ,  $x^{(i)}$  denotes its i-th component. The minimum between two reals is denoted by  $\wedge$ . We call  $I_2$  the identity matrix of dimension 2 and  $A^T$  is the transpose matrix of a matrix A. Finally, we denote by C some positive constant, which may change from line to line, and we use subscripts to indicate the parameters on which it depends when it is necessary.

## 2 Study of the system driven by a Brownian motion

In this section, the driving process L is supposed to be a standard Brownian motion, i.e.  $\alpha = 2$ . It will be denoted by B to keep standard notations. To be precise, (SKE) becomes

$$\begin{cases}
dV_t = dB_t - \operatorname{sgn}(V_t) \frac{|V_t|^{\gamma}}{t^{\beta}} dt - X_t dt, \\
dX_t = V_t dt, \\
(V_{t_0}, X_{t_0}) = (v_0, x_0).
\end{cases}$$
(3)

The previous system can be written in a vector viewpoint. Indeed, we set, for all  $t \geq t_0$  and  $v \in \mathbb{R}$ ,

$$Z_t := \begin{pmatrix} X_t \\ V_t \end{pmatrix}, \ W_t := \begin{pmatrix} 0 \\ B_t \end{pmatrix}, \ A := \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, \ \Gamma := \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \text{ and } F(t,v) := \begin{pmatrix} 0 \\ \operatorname{sgn}(v) \frac{|v|^{\gamma}}{t^{\beta}} \end{pmatrix}.$$

Thereby, the system (SKE) can be rewritten as

$$\begin{cases} dZ_t = \Gamma dW_t + AZ_t dt - F(t, V_t) dt, \\ Z_{t_0} = z_0 := (x_0, v_0)^T. \end{cases}$$
(4)

Notice that the matrix A is the rotation matrix of angle  $\frac{\pi}{2}$  and that, for all  $t \in \mathbb{R}$ ,

$$\Theta_t^{-1} := e^{tA} = \begin{pmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{pmatrix}.$$

We also define, for any  $t \ge t_0$ ,  $Y_t := e^{-tA}Z_t$ . We easily check, with Itô's formula, that Y is given by

$$dY_t = e^{-tA} \Gamma dW_t - e^{-tA} F(t, V_t) dt.$$
 (SDE<sub>Y</sub>)

## 2.1 Existence up to explosion

**Theorem 2.1.** The system of SDEs (3) admits a unique (global) strong solution if  $\gamma \in (0,1]$ . And if  $\gamma > 1$ , there exists a unique strong solution defined up to its explosion time  $\tau_{\infty}$ .

*Proof.* In the case  $\gamma > 1$ , the coefficients of the SDE (4) are locally Lipschitz continuous with respect to the space variable, locally uniformly in time. Thus, by Theorem 21.3 p. 415 in [Kal02], there exists a unique solution up to explosion. The argument is standard, see [SG03] for details. Assume now that  $\gamma \leq 1$ . We will use Theorem 1 in [HMC18]. Keeping the same notations, we

have in our case, for any  $(x,v) \in \mathbb{R}^2$  and  $t \geq t_0$ ,  $F_1(t,v,x) := -\operatorname{sgn}(v)|v|^{\gamma}t^{-\beta} - x$ ,  $F_2(t,v,x) := v$  and  $\sigma(t,v,x) = 1$ . Assumptions (ML) and (UE) in [HMC18] are obviously satisfied. Let us now remark that  $F_1$  is  $\gamma$ -Hölder with respect to  $v \in \mathbb{R}$  uniformly with respect to  $t \geq t_0$  and  $x \in \mathbb{R}$ , and is Lipschitz continuous with respect to x, uniformly with respect to t and v. With the notations used in [HMC18], we have  $\beta_1 = \gamma$  and  $\beta_2 = 1$ . Thus, Assumption ( $T_{\beta}$ ) is satisfied. Finally, we check that Assumption ( $H_{\eta}$ ) is satisfied. Since  $\partial_v F_2 = 1$ , we can conclude, taking  $\eta$  small enough and  $\mathcal{E}_1 = \{1\}$ , that there exists a unique strong solution to (3).

## 2.2 Moment estimates and non-explosion

In this section, we state and prove moment estimates of Z. It will be useful to control some stochastic terms appearing later. For all  $n \ge 0$ , define the stopping time

$$\tau_n := \inf\{t \ge t_0, \|Z_t\| \ge n\}.$$

Set  $\tau_{\infty} := \lim_{n \to +\infty} \tau_n$  the explosion time of Z.

**Proposition 2.2.** The explosion time of Z is a.s. infinite and, for all  $\kappa \geq 0$ ,

$$\mathbb{E}\left[\left\|Z_{t}\right\|^{\kappa}\right] \leq C_{\kappa,t_{0}} t^{\frac{\kappa}{2}}.\tag{5}$$

**Remark 2.3.** Let us mention that the moment estimate obtained for the position process X is a priori smaller in our case than in the free potential case [GL21a]. It is explained by the confining effect of the quadratic potential.

*Proof.* The proof is adapted from [GL21a] to two-dimensional processes. For the sake of completeness, we sketch the proof in our context.

Using Itô's formula applied to the function  $f:(x,v)\mapsto x^2+v^2$  and the fact that for all  $z\in\mathbb{R}^2$ ,  $z\cdot Az=0$ , we deduce that, for all  $t\geq t_0$ ,

$$||Z_{t \wedge \tau_n}||^2 \le ||z_0||^2 + \int_{t_0}^t 2\mathbb{1}_{\{s \le \tau_n\}} Z_s \cdot (\Gamma \, dW_s) - \int_{t_0}^{t \wedge \tau_n} 2Z_s \cdot F(s, V_s) \, ds + (t - t_0).$$

Remark that for any  $s \ge t_0$ ,  $Z_s \cdot F(s, V_s) = V_s \operatorname{sgn}(V_s) |V_s|^{\gamma} s^{-\beta} \ge 0$ . Taking expectation yields

$$\mathbb{E}\left[\|Z_{t \wedge \tau_n}\|^2\right] \le \|z_0\|^2 + (t - t_0) \le C_{t_0}t.$$

Thanks to Lemma B.1, we can conclude that the explosion time of Z is a.s. infinite. Set  $\kappa \in [0, 2]$ , so, by Jensen's inequality and Fatou's lemma

$$\mathbb{E}\left[\left\|Z_{t\wedge\tau_{\infty}}\right\|^{\kappa}\right] \leq \mathbb{E}\left[\left\|Z_{t\wedge\tau_{\infty}}\right\|^{2}\right]^{\frac{\kappa}{2}} \leq \left(\liminf_{n\to\infty} \mathbb{E}\left[\left\|Z_{t\wedge\tau_{n}}\right\|^{2}\right]\right)^{\frac{\kappa}{2}} \leq C_{\kappa,t_{0}} t^{\frac{\kappa}{2}}.$$
(6)

This leads to (5).

When  $\kappa > 2$ ,  $v \mapsto ||v||^{\kappa}$  is a  $\mathcal{C}^2$ -function, so by Itô's formula, for all  $t \geq t_0$ ,

$$||Z_{t \wedge \tau_{n}}||^{\kappa} \leq ||z_{0}||^{\kappa} + \int_{t_{0}}^{t \wedge \tau_{n}} \kappa ||Z_{s}||^{\kappa - 2} Z_{s} \cdot (\Gamma dW_{s}) - \int_{t_{0}}^{t \wedge \tau_{n}} \kappa ||Z_{s}||^{\kappa - 2} Z_{s} \cdot F(s, V_{s}) ds + \int_{t_{0}}^{t \wedge \tau_{n}} C_{\kappa} ||Z_{s}||^{\kappa - 2} ds.$$

In addition, it follows from the hypothesis on the sign of the drift function that

$$\|Z_{t \wedge \tau_n}\|^{\kappa} \le \|z_0\|^{\kappa} + \int_{t_0}^{t} \kappa \mathbb{1}_{\{s \le \tau_n\}} \|Z_s\|^{\kappa - 2} Z_s \cdot (\Gamma \, dW_s) + \int_{t_0}^{t \wedge \tau_n} C_{\kappa} \|Z_s\|^{\kappa - 2} \, ds. \tag{7}$$

Taking expectation in (7), we have

$$\mathbb{E}\left[\left\|Z_{t\wedge\tau_{\infty}}\right\|^{\kappa}\right] \leq \liminf_{n\to\infty} \mathbb{E}\left[\left\|Z_{t\wedge\tau_{n}}\right\|^{\kappa}\right] \leq \left\|v_{0}\right\|^{\kappa} + \int_{t_{0}}^{t} C_{\kappa} \mathbb{E}\left[\left\|Z_{s}\right\|^{\kappa-2}\right] \mathrm{d}s.$$

When  $0 \le \kappa - 2 \le 2$ , we can upper bound  $\mathbb{E}\left[\|Z_s\|^{\kappa - 2}\right]$  by injecting (6) and get

$$\mathbb{E}\left[\left\|Z_{t\wedge\tau_{\infty}}\right\|^{\kappa}\right] \leq \left\|v_{0}\right\|^{\kappa} + \int_{t_{0}}^{t} C_{\kappa,t_{0}} s^{\frac{\kappa-2}{2}} \, \mathrm{d}s \leq C_{\kappa,t_{0}} s^{\frac{\kappa}{2}}.$$

The method is then applied inductively to prove the inequality for all  $\kappa > 2$ .

## 2.3 Asymptotic behavior of the solution

We gather in this section the proof of Theorem 1.1. The strategy is to prove the convergence of the finite dimensional distributions (f.d.d.) of the process  $Y^{(\varepsilon)}$ , as  $\varepsilon \to 0$ , and its tightness in the critical and super-critical regimes. We first focus on the tightness.

**Lemma 2.4.** If  $2q \ge 1$ , then the family  $\{(\sqrt{\varepsilon}Y_{t/\varepsilon})_{t \ge \varepsilon t_0}, \ \varepsilon > 0\}$  is tight on every compact interval [m, M], with  $0 < m \le M$ .

*Proof.* We use the Kolmogorov criterion stated in Problem 4.11 p. 64 in [KS98]. Take  $\varepsilon_0$  small enough such that for all  $\varepsilon \leq \varepsilon_0$ , we have  $\varepsilon t_0 \leq m$ . Fix  $m \leq s \leq t \leq M$  and a > 4. Define, for  $t \geq \varepsilon t_0$ , the local martingale term appearing in  $(SDE_Y)$ 

$$M_t^{(\varepsilon)} := \sqrt{\varepsilon} \int_{t_0}^{t/\varepsilon} e^{-sA} \Gamma \, dW_s = \sqrt{\varepsilon} \int_{t_0}^{t/\varepsilon} \begin{pmatrix} -\sin(s) \\ \cos(s) \end{pmatrix} dB_s. \tag{8}$$

Using Jensen's inequality, moment estimates (see Proposition 2.2) and Burkholder-Davis-Gundy's inequality (see Theorem 4.4.22 p. 263 in [App09]), we have

$$\mathbb{E}\left[\left\|Y_{t}^{(\varepsilon)} - Y_{s}^{(\varepsilon)}\right\|^{a}\right] \leq C_{a}\mathbb{E}\left[\left\|M_{t}^{(\varepsilon)} - M_{s}^{(\varepsilon)}\right\|^{a}\right] + C_{a}\mathbb{E}\left[\left\|\sqrt{\varepsilon}\int_{s/\varepsilon}^{t/\varepsilon}e^{-uA}F(u, V_{u})\,\mathrm{d}u\right\|^{a}\right]$$

$$\leq C_{a}\mathbb{E}\left[\left\|M_{t}^{(\varepsilon)} - M_{s}^{(\varepsilon)}\right\|^{a}\right] + C_{a}\varepsilon^{1-\frac{a}{2}}(t-s)^{a-1}\mathbb{E}\left[\int_{s/\varepsilon}^{t/\varepsilon}\|F(u, V_{u})\|^{a}\,\mathrm{d}u\right]$$

$$\leq C_{a}\mathbb{E}\left[\left(\mathrm{Tr}\left(\left\langle M_{\cdot}^{(\varepsilon)} - M_{s}^{(\varepsilon)}\right\rangle_{t}\right)\right)^{a/2}\right] + C_{a}\varepsilon^{1-\frac{a}{2}}(t-s)^{a-1}\int_{s/\varepsilon}^{t/\varepsilon}u^{\frac{\gamma a}{2}-\beta a}\,\mathrm{d}u$$

$$\leq C_{a}(t-s)^{\frac{a}{2}} + C_{a,m,M}\varepsilon^{a(\beta-\frac{\gamma+1}{2})}(t-s)^{a-1}$$

$$\leq C_{a,m,M}(t-s)^{\frac{a}{2}}.$$

Since  $\frac{a}{2} > 2$  and  $\beta \ge \frac{\gamma+1}{2}$  the upper bound is independent of  $\varepsilon \le 1$ . Furthermore, by moment estimates (Proposition 2.2),

$$\sup_{\varepsilon < \varepsilon_0} \mathbb{E} \left[ \left\| Y_m^{(\varepsilon)} \right\| \right] \leq \sqrt{m} < \infty.$$

Thus, Kolmogorov's criterion can be applied, proving the tightness result.

We will now prove the convergence of the finite-dimensional distributions of  $Y^{(\varepsilon)}$ . Thanks to the previous lemma, this will yield the weak convergence on every compact set (see Theorem 13.1 p. 139 in [Bil99]). The convergence in distribution on the whole space  $\mathcal{C}$  will follow, for  $2q \ge 1$ , from Theorem 16.7 p. 174 in [Bil99], since all processes considered are continuous.

### 2.3.1 Convergence of the f.d.d. in the super-critical regime

Assume here that 2q > 1. Recall that  $(Y_t^{(\varepsilon)})_{t > \varepsilon t_0} := (\sqrt{\varepsilon} Y_{t/\varepsilon})_{t > \varepsilon t_0}$ .

Proof of Theorem 1.1 (i).

Step 1. We first prove the convergence of the f.d.d. of the local martingale term  $M^{(\varepsilon)} =$  $(M^{(\varepsilon,1)},M^{(\varepsilon,2)})^T$  appearing in  $(SDE_Y)$ . Recall that the stochastic integral  $M^{(\varepsilon)}$  was defined in (8). It is a centered Gaussian process

with covariance kernel defined, for any  $(s,t) \in [\varepsilon t_0, +\infty)^2$ , by

$$K^{(\varepsilon)}(s,t) := \begin{pmatrix} \operatorname{Cov}(M_s^{(\varepsilon)}) & \operatorname{Cov}(M_s^{(\varepsilon)}, M_t^{(\varepsilon)}) \\ \operatorname{Cov}(M_t^{(\varepsilon)}, M_s^{(\varepsilon)}) & \operatorname{Cov}(M_t^{(\varepsilon)}) \end{pmatrix},$$

where

$$\mathrm{Cov}(M_s^{(\varepsilon)}, M_t^{(\varepsilon)}) = \begin{pmatrix} \mathrm{Cov}(M_s^{(\varepsilon,1)}, M_t^{(\varepsilon,1)}) & \mathrm{Cov}(M_s^{(\varepsilon,1)}, M_t^{(\varepsilon,2)}) \\ \mathrm{Cov}(M_s^{(\varepsilon,2)}, M_t^{(\varepsilon,1)}) & \mathrm{Cov}(M_s^{(\varepsilon,2)}, M_t^{(\varepsilon,2)}) \end{pmatrix},$$

and  $\operatorname{Cov}(M_s^{(\varepsilon)}) = \operatorname{Cov}(M_s^{(\varepsilon)}, M_s^{(\varepsilon)})$ . Thus, the convergence of the f.d.d. of  $M^{(\varepsilon)}$  reduces on the study of the limit of  $K^{(\varepsilon)}$ , when  $\varepsilon$  converges to 0. Let us fix  $\varepsilon t_0 \leq s \leq t$ . Using that  $M^{(\varepsilon)}$  has independent increments and by Itô's isometry, we find that

$$\mathrm{Cov}(M_s^{(\varepsilon)}, M_t^{(\varepsilon)}) = \begin{pmatrix} \varepsilon \int_{t_0}^{s/\varepsilon} \sin(u)^2 \, \mathrm{d}u & -\varepsilon \int_{t_0}^{s/\varepsilon} \sin(u) \cos(u) \, \mathrm{d}u \\ -\varepsilon \int_{t_0}^{s/\varepsilon} \sin(u) \cos(u) \, \mathrm{d}u & \varepsilon \int_{t_0}^{s/\varepsilon} \cos(u)^2 \, \mathrm{d}u \end{pmatrix}.$$

We get that, for all  $0 < s \le t$ ,

$$\mathrm{Cov}(M_s^{(\varepsilon)}, M_t^{(\varepsilon)}) \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad \frac{1}{2} \begin{pmatrix} s & 0 \\ 0 & s \end{pmatrix}.$$

We recognize the covariance kernel of the process  $\left(\mathcal{B}_{\frac{t}{2}}\right)_{t>0}$ , where  $\mathcal{B}$  denotes a standard Brownian motion on  $\mathbb{R}^2$ . Since mean and covariance functions characterize Gaussian process (see Lemma 13.1 (i) p. 250 in [Kal02]), we have thus proved that  $(M_t^{(\varepsilon)})_{t \geq \varepsilon t_0}$  converges in f.d.d. towards  $\left(\mathcal{B}_{\frac{t}{2}}\right)_{t>0}$ .

Step 2. Pick T > 0. We prove that

$$\mathbb{E}\left[\sup_{\varepsilon t_0 < t < T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \right] \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad 0.$$

We have

$$\sup_{\varepsilon t_0 \le t \le T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \le \sqrt{\varepsilon} \|z_0\| + \sqrt{\varepsilon} \int_{t_0}^{T/\varepsilon} \left\| e^{-sA} F(s, V_s) \right\| \mathrm{d}s.$$

We use moment estimates (Proposition 2.2) to get

$$\mathbb{E}\left[\sqrt{\varepsilon} \int_{t_0}^{T/\varepsilon} \left\| e^{-sA} F(s, V_s) \right\| \mathrm{d}s \right] = \mathbb{E}\left[\sqrt{\varepsilon} \int_{t_0}^{T/\varepsilon} \left\| F(s, V_s) \right\| \mathrm{d}s \right] \\
\leq \mathbb{E}\left[\sqrt{\varepsilon} \int_{t_0}^{T/\varepsilon} \left| V_s \right|^{\gamma} s^{-\beta} \mathrm{d}s \right] \\
\leq \sqrt{\varepsilon} C_{\kappa, t_0} \int_{t_0}^{T/\varepsilon} s^{\frac{\gamma}{2} - \beta} \mathrm{d}s \\
\leq C_{\kappa, t_0} \left(\varepsilon^{\beta - \frac{\gamma + 1}{2}} T^{\frac{\gamma}{2} - \beta + 1} - \sqrt{\varepsilon} t_0^{\frac{\gamma}{2} - \beta + 1} \right).$$

Hence, setting  $r := \min(\beta - \frac{\gamma+1}{2}, \frac{1}{2})$ , which is positive by assumption, we get

$$\mathbb{E}\left[\sup_{\varepsilon t_0 \leq t \leq T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \right] = \mathop{O}_{\varepsilon \to 0}(\varepsilon^r).$$

We conclude the proof using Theorem 3.1 p. 27 in [Bil99].

#### 2.3.2 Convergence of the f.d.d. in the critical and sub-critical regimes

In this section, we consider the linear case, i.e.  $\gamma = 1$ . Pick  $\beta \in (\frac{1}{2}, 1]$ . Recall that

$$(Y_t^{(\varepsilon)})_{t \ge \varepsilon t_0} := (\varepsilon^q Y_{t/\varepsilon})_{t \ge \varepsilon t_0},$$

where  $q = \frac{\beta}{\gamma + 1}$ .

*Proof of Theorem 1.1 (ii) and (iii)*. Leaving out the Brownian term, the underlying ODE of our system is the following

$$x''(t) + \frac{x'(t)}{t^{\beta}} + x(t) = 0, \quad t \ge t_0.$$
 (9)

Pick the basis of solutions given in Lemma A.3 and denote by R its resolvent matrix. By Itô's formula we get that for all  $t \ge t_0$ ,

$$R_t^{-1} Z_t = R_{t_0}^{-1} Z_0 + \int_{t_0}^t R_s^{-1} \Gamma \, \mathrm{d}W_s.$$

Let us define f the rate of decrease of R (see Lemma A.3) by

$$\forall t > 0, f(t) := \begin{cases} \frac{1}{\sqrt{t}} & \text{if } \beta = 1, \\ \exp\left(-\frac{t^{1-\beta}}{2(1-\beta)}\right) & \text{else.} \end{cases}$$

Set, for  $t \geq \varepsilon t_0$ ,

$$\Phi_t := \frac{e^{-tA}R_t}{f(t)} \quad \text{and} \quad \widetilde{M}_t^{(\varepsilon)} := \varepsilon^q f\left(\frac{t}{\varepsilon}\right) \int_{t_0}^{t/\varepsilon} R_s^{-1} \Gamma \, \mathrm{d}W_s. \tag{10}$$

Pick  $t \geq \varepsilon t_0$ . To study the convergence of  $Y^{(\varepsilon)}$  we decompose it into

$$Y_t^{(\varepsilon)} = \varepsilon^q f\left(\frac{t}{\varepsilon}\right) \Phi_{t/\varepsilon} R_{t_0}^{-1} Z_0 + \Phi_{t/\varepsilon} \widetilde{M}_t^{(\varepsilon)}. \tag{11}$$

Let us notice that, for any t > 0,  $\varepsilon^q f\left(\frac{t}{\varepsilon}\right)$  converges to 0, as  $\varepsilon \to 0$ .

Step 1. We first note the convergence of  $\Phi$ .

Using the asymptotic expansion of the resolvent matrix (Lemma A.3), we can write, for  $t \geq \varepsilon t_0$ ,

$$\Phi_t = I_2 + \mathop{O}_{t \to \infty} \left( t^{1-2\beta} \right).$$

As a consequence, since  $1 - 2\beta < 0$ ,  $\Phi_{t/\varepsilon}$  converges to the identity matrix  $I_2$ , as  $\varepsilon \to 0$ . Moreover, we obtain that

 $\varepsilon^q f\left(\frac{t}{\varepsilon}\right) \Phi_{t/\varepsilon} R_{t_0}^{-1} Z_0 \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad 0.$ 

Therefore, we can forget the first term appearing in the decomposition (11) of  $Y^{(\varepsilon)}$  (see Theorem 3.1 p. 27 in [Bil99]).

STEP 2. We compute now the covariance kernel of  $Y^{(\varepsilon)}$  defined in (11). It is defined, for  $(s,t) \in [\varepsilon t_0, +\infty)^2$ , by

$$K^{(\varepsilon)}(s,t) := \begin{pmatrix} \operatorname{Cov}(Y_s^{(\varepsilon)}) & \operatorname{Cov}(Y_s^{(\varepsilon)},Y_t^{(\varepsilon)}) \\ \operatorname{Cov}(Y_t^{(\varepsilon)},Y_s^{(\varepsilon)}) & \operatorname{Cov}(Y_t^{(\varepsilon)}) \end{pmatrix},$$

where

$$\operatorname{Cov}(Y_s^{(\varepsilon)},Y_t^{(\varepsilon)}) = \begin{pmatrix} \operatorname{Cov}(Y_s^{(\varepsilon,1)},Y_t^{(\varepsilon,1)}) & \operatorname{Cov}(Y_s^{(\varepsilon,1)},Y_t^{(\varepsilon,2)}) \\ \operatorname{Cov}(Y_s^{(\varepsilon,2)},Y_t^{(\varepsilon,1)}) & \operatorname{Cov}(Y_s^{(\varepsilon,2)},Y_t^{(\varepsilon,2)}) \end{pmatrix},$$

and  $Cov(Y_s^{(\varepsilon)}) = Cov(Y_s^{(\varepsilon)}, Y_s^{(\varepsilon)}).$ 

Moreover, we get, for all  $(s,t) \in [\varepsilon t_0, +\infty)^2$ 

$$\operatorname{Cov}\left(\Phi_{s/\varepsilon}\widetilde{M}_{s}^{(\varepsilon)},\Phi_{t/\varepsilon}\widetilde{M}_{t}^{(\varepsilon)}\right) = \Phi_{s/\varepsilon}\operatorname{Cov}(\widetilde{M}_{s}^{(\varepsilon)},\widetilde{M}_{t}^{(\varepsilon)})\Phi_{t/\varepsilon}^{T}.$$

Using the expression of the Wronskian obtained in Lemma A.3, we obtain, for all  $t \geq \varepsilon t_0$ ,

$$\widetilde{M}_t^{(\varepsilon)} = \varepsilon^q f(t/\varepsilon) \int_{t_0}^{t/\varepsilon} f(u)^{-2} \begin{pmatrix} -y_2(u) \\ y_1(u) \end{pmatrix} dB_u.$$

It is a centered Gaussian process and for any  $\varepsilon t_0 \leq s \leq t$ , we have

$$\operatorname{Cov}(\widetilde{M}_{s}^{(\varepsilon)}, \widetilde{M}_{t}^{(\varepsilon)}) = \varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_{0}}^{s/\varepsilon} f(u)^{-4} \begin{pmatrix} y_{2}^{2}(u) & -y_{2}(u)y_{1}(u) \\ -y_{2}(u)y_{1}(u) & y_{1}^{2}(u) \end{pmatrix} du.$$

Using the asymptotic expansion of the solutions and Lemma B.3, we get, for all  $\varepsilon t_0 < s \le t$ ,

$$\operatorname{Cov}(\widetilde{M}_{s}^{(\varepsilon)}, \widetilde{M}_{t}^{(\varepsilon)}) = \varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_{0}}^{s/\varepsilon} f(u)^{-2} \begin{pmatrix} \sin^{2}(u) & -\sin(u)\cos(u) \\ -\sin(u)\cos(u) & \cos^{2}(u) \end{pmatrix} du + \underset{\varepsilon \to 0}{O} \left( \varepsilon^{2\beta-1} f(t/\varepsilon) f(s/\varepsilon)^{-1} \right).$$

Moreover, using asymptotic expansions of these integrals (see Lemmas B.2 and B.3),

$$\varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_0}^{s/\varepsilon} f(u)^{-2} \cos^2(u) \, du = \varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \frac{1}{2} \int_{t_0}^{s/\varepsilon} f(u)^{-2} \, du + \underset{\varepsilon \to 0}{o} \left( f(t/\varepsilon) f(s/\varepsilon)^{-1} \right).$$

The same equality holds for

$$\varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_0}^{s/\varepsilon} f(u)^{-2} \sin^2(u) du,$$

and we have

$$\varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_0}^{s/\varepsilon} f(u)^{-2} \cos(u) \sin(u) du = \mathop{o}_{\varepsilon \to 0} \left( f(t/\varepsilon) f(s/\varepsilon)^{-1} \right).$$

Thanks to Lemma B.3, this leads to

$$\operatorname{Cov}(\widetilde{M}_{s}^{(\varepsilon)}, \widetilde{M}_{t}^{(\varepsilon)}) = \left[\frac{1}{2}\varepsilon^{\beta} f(t/\varepsilon) f(s/\varepsilon) \int_{t_{0}}^{s/\varepsilon} f(u)^{-2} du\right] I_{2} + \underset{\varepsilon \to 0}{o} \left(f(t/\varepsilon) f(s/\varepsilon)^{-1}\right)$$
$$= k_{\beta} \frac{f(t/\varepsilon)}{f(s/\varepsilon)} s^{\beta} I_{2} + \underset{\varepsilon \to 0}{o} \left(f(t/\varepsilon) f(s/\varepsilon)^{-1}\right),$$

where

$$k_{\beta} := \begin{cases} \frac{1}{4} & \text{if } \beta = 1, \\ \frac{1}{2} & \text{else.} \end{cases}$$

Thus, we have proved the convergence of the f.d.d. of  $Y^{(\varepsilon)}$ . Note that whenever 2q=1, we recognize the covariance kernel of the process  $\left(\frac{1}{\sqrt{2t}}\int_0^t \sqrt{s}\,\mathrm{d}\mathcal{B}_s\right)_{t>0}$ , where  $\mathcal{B}$  denotes a standard Brownian motion on  $\mathbb{R}^2$ .

- Remark 2.5. The proof relies on the asymptotic expansion of the resolvent matrix of (9). We were able to prove it only for  $\beta \in \left(\frac{1}{2},1\right]$ . However, if  $\beta = 0$ , the resolvent matrix is explicit and following the same lines, we can prove that  $\left(Z_{t/\varepsilon}\right)_{t \geq \varepsilon t_0}$  converges in f.d.d. towards a centered Gaussian process with covariance kernel  $(s,t) \mapsto \frac{1}{2}I_2\mathbb{1}_{\{s=t\}}$ . This behavior can be explained by the fact that the frictional force does not decrease along time. This cancels somehow the rotation bearing, which prevents  $Z^{(\varepsilon)}$  from converging as a process when  $\beta > 0$ .
  - The asymptotic expansion of the resolvant matrix is also konwn in the super-critical regime, i.e.  $\beta > 1$ . Therefore, one can prove the result in the linear case, i.e.  $\gamma = 1$ , following the same lines.

## 2.4 Proof of Corollary 1.5

Proof of Corollary 1.5. We start by proving the convergence in distribution of  $v_{1/T}Z_T$ , as  $T \to +\infty$ . We claim that it follows from Theorem 1.1. Indeed, it is enough to remark that the convergence results stated in Theorem 1.1 imply the convergence in distribution of the marginal distribution at time t=1 of  $Y^{(\varepsilon)}$ . Let us also recall that  $Z_T=e^{TA}Y_T$ . Setting  $T=\frac{1}{\varepsilon}$ , the convergence of  $v_{1/T}Z_T$  is therefore a direct consequence of Lemma B.4.

We now show that the rescaled process  $Z^{(\varepsilon)}$  does not converge in distribution. We do the proof only in the super-critical regime. Assume by contradiction that it is the case. Hence, each of

its coordinates shall converge too. We thus have the convergence of the rescaled process  $X^{(\varepsilon)}$ . Using  $(SDE_Y)$ , we can write

$$\sqrt{\varepsilon}X_{t/\varepsilon} = \sqrt{\varepsilon}x_0 + \sqrt{\varepsilon}\int_{t_0}^{t/\varepsilon} \sin\left(\frac{t}{\varepsilon} - s\right) dB_s - \sqrt{\varepsilon}\int_{t_0}^{t/\varepsilon} \sin\left(\frac{t}{\varepsilon} - s\right) F(V_s) s^{-\beta} ds.$$

As in the proof of Theorem 1.1, the last term converges in probability uniformly on compact intervals towards zero. Hence, the following term shall converge in distribution

$$I_t^{(\varepsilon)} := \sqrt{\varepsilon} \int_{t_0}^{t/\varepsilon} \sin\left(\frac{t}{\varepsilon} - s\right) \mathrm{d}B_s.$$

The process  $(I_t^{(\varepsilon)})_{t \geq \varepsilon t_0}$  is Gaussian, thereby its limit shall be Gaussian too and its covariance function shall converge (see Lemma 13.1 (i) p. 250 in [Kal02]). However, using Itô's isometry, one can compute, for  $\varepsilon t_0 \leq s \leq t$ ,

$$\mathbb{E}\left[I_t^{(\varepsilon)}I_s^{(\varepsilon)}\right] = \varepsilon \int_{t_0}^{s/\varepsilon} \sin\left(\frac{t}{\varepsilon} - u\right) \sin\left(\frac{s}{\varepsilon} - u\right) du$$

$$= \varepsilon \frac{1}{2} \left[\cos\left(\frac{t - s}{\varepsilon}\right) \left(\frac{s}{\varepsilon} - t_0\right) + \frac{1}{2} \left(\sin\left(\frac{t - s}{\varepsilon}\right) - \sin\left(\frac{t - s}{\varepsilon} - 2t_0\right)\right)\right]$$

$$= \frac{1}{2} s \cos\left(\frac{t - s}{\varepsilon}\right) + \underset{\varepsilon \to 0}{o}(1).$$

This term does not converge if  $s \neq t$ , and that concludes the proof.

## 3 Study of the system driven by an $\alpha$ -stable process

In this section, L is a symmetric  $\alpha$ -stable Lévy process. We call  $\nu$  its Lévy measure, which can be written as  $\nu(\mathrm{d}z) = a |z|^{-1-\alpha} \mathbb{1}_{\{z\neq 0\}} \, \mathrm{d}z$  with a>0. As a Lévy measure, it satisfies  $\int_{\mathbb{R}^*} (1 \wedge z^2) \nu(\mathrm{d}z) < +\infty$ . We denote by N the Poisson random measure associated with L and by  $\widetilde{N}$  its compensated Poisson measure. Using Lévy-Itô's decomposition, we have, for all  $t\geq 0$ ,

$$L_t = \int_0^t \int_{\mathbb{R}^*} z \widetilde{N}(\mathrm{d}s, \mathrm{d}z)$$

As in the previous section, we set, for all  $t \geq t_0$  and  $v \in \mathbb{R}$ ,

$$Z_t := \begin{pmatrix} X_t \\ V_t \end{pmatrix}, \ S_t := \begin{pmatrix} 0 \\ L_t \end{pmatrix}, \ A := \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, \ \Gamma := \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \text{ and } F(t, v) := \begin{pmatrix} 0 \\ \operatorname{sgn}(v) \frac{|v|^{\gamma}}{t^{\beta}} \end{pmatrix}.$$

Thereby, the system (SKE) can be rewritten as

$$\begin{cases} dZ_t = \Gamma dS_t + AZ_t dt - F(t, V_t) dt, \\ Z_{t_0} = z_0 := (x_0, v_0)^T. \end{cases}$$
(12)

We define, for any  $t \ge t_0$ ,  $Y_t := e^{-tA}Z_t$ . We easily check, with Itô's formula, that Y is given by

$$dY_t = e^{-tA} \Gamma dS_t - e^{-tA} F(t, V_t) dt.$$
 (SDE<sub>Y</sub>)

## 3.1 Existence up to explosion

**Theorem 3.1.** The system (SKE) admits a unique weak solution if  $\gamma \in (0,1]$ . If  $\gamma > 1$ , there exists a unique strong solution defined up to its explosion time  $\tau_{\infty}$ .

*Proof.* In the case  $\gamma > 1$ , the coefficients of the SDE (12) satisfied by Z = (X,V) are locally Lipschitz continuous with respect to the space variable, locally uniformly in time. So we can apply Lemma 115 p. 78 in [Sit05] to get the pathwise uniqueness. The drift coefficient is continuous with respect to its two variables, so it is a locally bounded and measurable function. By a standard localization argument, using Theorem 9.1 p. 231 in [IW81], since the drift coefficient is locally Lipschitz, there is a unique solution defined up to explosion.

Assume now that  $\gamma \leq 1$ . We check that we can use Theorem 1 in [MM21]. Using the same notations, we have

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

and for any  $(t, x_1, x_2) \in [t_0, +\infty) \times \mathbb{R}^2$ ,  $F_1(t, x_1, x_2) = \operatorname{sgn}(x_1)|x_1|^{\gamma}t^{-\beta}$ ,  $F_2(t, x_1, x_2) = 0$  and  $\sigma(t, x_1, x_2) = 1$ . Assumptions **(UE)** and **(ND)** are clearly satisfied. Since  $F_2$  does not depend on  $x_1$  and since  $[\widetilde{A}]_{2,1} = 1$  is different from 0, we deduce that Assumption **(H)** is satisfied. We easily check that Theorem 1 in [MM21] can be applied with  $\beta_1 = \gamma$ , and  $\beta_2 = 1$ .

**Remark 3.2.** For  $\alpha \in (0,2)$ , employing the technique of Picard iteration and the interlacing procedure, one can deduce that (12) has a unique solution in the linear setting  $\gamma = 1$  (see [App09, p. 375]).

## 3.2 Moment estimates and non-explosion

Let Z be the unique solution up to explosion time to (12). As in the continuous setting, define, for all  $r \ge 0$ , the stopping time

$$\tau_r := \inf\{t > t_0, \|Z_t\| > r\}.$$

Set  $\tau_{\infty} := \lim_{r \to +\infty} \tau_r$  the explosion time of Z. For the sake of simplicity, since there is no jump on the position component, for  $z \in \mathbb{R}$ , we shall write  $Z_{s-} + z$  for  $(X_s, V_{s-} + z)$  in the following. We adapt the proof of [GL21b] to two-dimensional processes.

**Proposition 3.3.** For any  $\gamma \geq 0$  and  $\beta \geq 0$ , the explosion time  $\tau_{\infty}$  is a.s. infinite and for  $\kappa \in (0, \alpha)$ , there exists  $C_{\kappa, t_0}$  such that

$$\forall t \ge t_0, \ \mathbb{E}\left[\|Z_t\|^{\kappa}\right] \le C_{\kappa, t_0} t^{\frac{\kappa}{\alpha}}. \tag{13}$$

**Remark 3.4.** Note that, as in the Brownian case, the moment estimates obtained for the position process X is a priori smaller in our case than in the free potential case [GL21b]. It is explained by the confining effect of the quadratic potential.

*Proof.* The key idea is to slice the small and big jumps in a non-homogeneous way with respect to the characteristic scale of an  $\alpha$ -stable process  $\xi \mapsto \xi^{\frac{1}{\alpha}}$ .

Pick  $\xi \geq t_0$ . The  $\alpha$ -stable symmetric Lévy driving process can be written as

$$L_t - L_{t_0} = \int_{t_0}^t \int_{|z| \le \xi^{\frac{1}{\alpha}}} z\widetilde{N}(\mathrm{d}s, \mathrm{d}z) + \int_{t_0}^t \int_{|z| > \xi^{\frac{1}{\alpha}}} zN(\mathrm{d}s, \mathrm{d}z).$$

STEP 1. We first apply Itô's formula (see Theorem 4.47 p. 251 in [App09]) and estimate the expectation of each term for  $\kappa \leq 1$ , in order to get (13).

Fix  $\eta > 0$  to be chosen latter and define the  $C^2$ -function  $f:(x,v) \mapsto (\eta + x^2 + v^2)^{\kappa/2}$ . We use the fact that for all  $y \in \mathbb{R}^2$ ,  $y \cdot Ay = 0$ , and observe that for any  $s \geq t_0$  and  $(x,v) \in \mathbb{R}^2$ ,  $(x,v)^T \cdot F(s,v) = |v|^{\gamma+1} s^{-\beta} \geq 0$ . For all  $t \geq t_0$ , by Itô's formula, we have

$$f(Z_{t \wedge \tau_r}) \le f(V_0) + M_t + R_t + S_t,$$

where

$$M_{t} := \int_{t_{0}}^{t} \int_{0 < |z| < \xi^{\frac{1}{\alpha}}} \mathbb{1}_{\{s \le \tau_{r}\}} (f(Z_{s-} + z) - f(Z_{s-})) \widetilde{N}(ds, dz),$$

$$R_{t} := \int_{t_{0}}^{t} \int_{|z| > \xi^{\frac{1}{\alpha}}} \mathbb{1}_{\{s \le \tau_{r}\}} (f(Z_{s-} + z) - f(Z_{s-})) N(ds, dz), \tag{14}$$

$$S_t := \int_{t_0}^t \int_{0 < |z| < \varepsilon^{\frac{1}{\alpha}}} \mathbb{1}_{\{s \le \tau_r\}} \left[ f(Z_{s-} + z) - f(Z_{s-}) - \mathrm{d}f(Z_{s-}).z \right] \nu(\mathrm{d}z) \,\mathrm{d}s. \tag{15}$$

Moreover, remark that for all  $k > \alpha$ .

$$\int_{0<|z|<\xi^{\frac{1}{\alpha}}} |z|^k \nu(\mathrm{d}z) = \frac{2a}{k-\alpha} \xi^{\frac{k}{\alpha}-1},\tag{16}$$

and for all  $k < \alpha$ ,

$$\int_{|z|>\xi^{\frac{1}{\alpha}}} |z|^k \nu(\mathrm{d}z) = \frac{2a}{\alpha - k} \xi^{\frac{k}{\alpha} - 1}.$$
 (17)

We estimate expectations of M, R and S.

To that end, we first show that the local martingale  $(M_t)_{t\geq t_0}$  is a martingale. Fix  $q\geq 2$  and  $r\geq 0$ . Moreover, we set

$$I_t(q) := \int_{t_0}^t \int_{0 < |z| < \varepsilon^{\frac{1}{\alpha}}} \mathbb{1}_{\{s \le \tau_r\}} |f(Z_{s-} + z) - f(Z_{s-})|^q \nu(\mathrm{d}z) \, \mathrm{d}s.$$

Thanks to Taylor-Lagrange inequality, for all  $\|(x,v)\| \leq r$  and  $|z| \leq \xi^{\frac{1}{\alpha}}$ ,

$$|f(x, v + z) - f(x, v)| \le \sup\{\|\nabla f(y)\|, \|y\| \in [-r - \xi^{1/\alpha}, r + \xi^{1/\alpha}]\} |z| \le C_{r,\xi,\kappa} |z|,$$

so we have

$$I_t(q) \le C_{r,\xi,\kappa} \int_{t_0}^t \int_{0<|z|<\xi^{\frac{1}{\alpha}}} \mathbb{1}_{\{s \le \tau_r\}} |z|^q \nu(\mathrm{d}z) \,\mathrm{d}s.$$

Hence, it is a finite quantity, since  $q \ge 2$  and (16) holds. Therefore, for  $q \ge 2$ , by Kunita's inequality (see Theorem 4.4.23 p. 265 in [App09]), there exists  $D_q > 0$  such that

$$\mathbb{E}\left[\sup_{t_0 \le s \le t} |M_s|^q\right] \le D_q\left(\mathbb{E}\left[I_t(2)^{\frac{q}{2}}\right] + \mathbb{E}\left[I_t(q)\right]\right) < +\infty.$$

Hence, by Theorem 51 p. 38 in [Pro05], M is a martingale.

We estimate now the finite variation part S defined in (15). Note that for all  $(x, v) \in \mathbb{R}^2$ , the Hessian matrix of f is given by

$$\operatorname{Hess}(f)(x,v) = \kappa(x^2 + v^2 + \eta)^{\frac{\kappa}{2} - 1} \begin{pmatrix} 1 + (\kappa - 2) \frac{x^2}{x^2 + v^2 + \eta} & (\kappa - 2) \frac{xv}{x^2 + v^2 + \eta} \\ (\kappa - 2) \frac{xv}{x^2 + v^2 + \eta} & 1 + (\kappa - 2) \frac{v^2}{x^2 + v^2 + \eta} \end{pmatrix}.$$

Its matrix norm is bounded by  $C_{\kappa} \eta^{\frac{\kappa}{2}-1}$ .

Assume that  $|z| < \xi^{\frac{1}{\alpha}}$ . Using Taylor- Lagrange's inequality and injecting (16) we get the almost sure following bound, for all  $s \ge t_0$ ,

$$\left| \int_{0 < |z| < \xi^{\frac{1}{\alpha}}} \left( f(Z_{s-} + z) - f(Z_{s-}) - \nabla f(Z_{s-}) \cdot z \right) \nu(\mathrm{d}z) \right| \le C_{\kappa} \eta^{\frac{\kappa}{2} - 1} \frac{2a}{2 - \alpha} \xi^{\frac{2}{\alpha} - 1}. \tag{18}$$

It remains to study the Poisson integral R defined in (14). Pick  $\kappa \leq 1$ , by Hölder property of power functions and (17), we deduce that

$$\int_{|z|>\xi^{\frac{1}{\alpha}}} |f(Z_{s-}+z) - f(Z_{s-})| \,\nu(\mathrm{d}z) \le \eta^{\frac{\kappa}{2}} \frac{2a}{\alpha} \xi^{-1} + \frac{2a}{\alpha - \kappa} \xi^{\frac{\kappa}{\alpha} - 1}. \tag{19}$$

Moment estimate of the Poisson integral follows from Theorem 2.3.7 p. 106 in [App09]. Gathering (19) and (18), we obtain

$$\mathbb{E}\left[\left\|Z_{t\wedge\tau_r}\right\|^{\kappa}\right] \leq \mathbb{E}\left[f(Z_{t\wedge\tau_r})\right] \leq \mathbb{E}\left[f(Z_{t_0})\right] + t\xi^{-1}\left(\eta^{\kappa/2}\frac{2a}{\alpha} + \frac{2a}{\alpha-\kappa}\xi^{\frac{\kappa}{\alpha}} + C_{\kappa}\eta^{\frac{\kappa}{2}-1}\frac{2a}{2-\alpha}\xi^{\frac{2}{\alpha}}\right).$$

Choosing  $\eta = t^{\frac{2}{\alpha}}$  and  $\xi = t$ , we get

$$\mathbb{E}\left[\left\|Z_{t \wedge \tau_r}\right\|^{\kappa}\right] \leq \mathbb{E}\left[f(Z_{t_0})\right] + t^{\frac{\kappa}{\alpha}} \left(\frac{2a}{\alpha} + \frac{2a}{\alpha - \kappa} + C_{\kappa} \frac{2a}{2 - \alpha}\right) \leq C_{\kappa, t_0} t^{\frac{\kappa}{\alpha}}.$$

Thanks to Lemma B.1, we can conclude that the explosion time of Z is a.s. infinite, and letting  $r \to +\infty$  with Fatou's lemma, for all  $\kappa \in [0, 1]$ ,

$$\mathbb{E}\left[\left\|Z_{t}\right\|^{\kappa}\right] \leq C_{\kappa,t_{0}} t^{\frac{\kappa}{\alpha}}.\tag{20}$$

STEP 2. Pick  $\kappa \in (1, \alpha)$ . We estimate R in another way, using again Theorem 2.3.7 p. 106 in [App09].

By the Hölder property of power function and (17), we get

$$\int_{|z| \ge \xi^{\frac{1}{\alpha}}} |f(Z_{s-} + z) - f(Z_{s-})| \nu(\mathrm{d}z) \le \int_{|z| \ge \xi^{\frac{1}{\alpha}}} |2zV_{s-} + z^2|^{\frac{\kappa}{2}} \nu(\mathrm{d}z) 
\le C_{\kappa} \left( \frac{2a}{\alpha - \kappa} \xi^{\frac{\kappa}{\alpha} - 1} + |V_{s-}|^{\frac{\kappa}{2}} \frac{2a}{\alpha - \frac{\kappa}{2}} \xi^{\frac{\kappa}{2\alpha} - 1} \right).$$
(21)

Gathering (18) and (21), one has

$$\mathbb{E}\left[\left\|Z_{t\wedge\tau_{r}}\right\|^{\kappa}\right] \leq \mathbb{E}\left[f(Z_{t_{0}})\right] + t\left(C_{\kappa}\frac{2a}{\alpha - \kappa}\xi^{\frac{\kappa}{\alpha} - 1} + C_{\kappa}\eta^{\frac{\kappa}{2} - 1}\frac{2a}{2 - \alpha}\xi^{\frac{2}{\alpha} - 1}\right) + C_{\kappa}\frac{2a}{\alpha - \frac{\kappa}{2}}\xi^{\frac{\kappa}{2\alpha} - 1}\int_{t_{0}}^{t}\mathbb{E}\left[\left|V_{s}\right|^{\frac{\kappa}{2}}\right]ds.$$

Injecting (20) applied with  $\frac{\kappa}{2}$ , choosing  $\eta = t^{\frac{2}{\alpha}}$  and  $\xi = t$ , we get

$$\mathbb{E}\left[\|Z_{t\wedge\tau_{\kappa}}\|^{\kappa}\right] \leq C_{\kappa,t_{0},\alpha}t^{\frac{\kappa}{\alpha}}.$$

The conclusion of the proof follows, letting  $r \to +\infty$ .

## 3.3 Asymptotic behavior of the solution

We gather in this section the proof of Theorem 1.2. The strategy is to prove the convergence of the f.d.d. of the process  $Y^{(\varepsilon)}$ , and then its tightness both in the super-critical and critical regimes. We first prove the tightness when  $\alpha q \geq 1$ . Recall that  $q = \frac{\beta}{\gamma + \alpha - 1}$ .

**Lemma 3.5.** Assume that  $\alpha q \geq 1$ , then the family  $\left\{ (\varepsilon^{\frac{1}{\alpha}} Y_{t/\varepsilon})_{t \geq \varepsilon t_0}, \ \varepsilon > 0 \right\}$  is tight on every compact interval [m, M], for  $0 < m \leq M$ .

*Proof.* We check the Aldous's tightness criterion stated in Theorem 16.10 p. 178 in [Bil99]. Let  $a, \eta, T$  be positive reals. Let  $\tau$  be a discrete stopping time with finite range  $\mathcal{T}$ , bounded by T and fix  $\delta > 0$  and  $\varepsilon > 0$  small enough to be chosen latter. Define, for  $t \geq \varepsilon t_0$ , the Wiener-Lévy integral appearing in  $(SDE_Y)$ 

$$M_t^{(\varepsilon)} := \varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{t/\varepsilon} e^{-sA} \Gamma \, \mathrm{d}S_s = \varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{t/\varepsilon} \begin{pmatrix} -\sin(s) \\ \cos(s) \end{pmatrix} \, \mathrm{d}L_s. \tag{22}$$

We use Jensen's inequality to get, for  $r = \frac{\alpha}{2}$ ,

$$\mathbb{E}\left[\left\|Y_{\tau+\delta}^{(\varepsilon)}-Y_{\tau}^{(\varepsilon)}\right\|^{r}\right] \leq \mathbb{E}\left[\left\|\int_{\tau/\varepsilon}^{(\tau+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_{u}\right\|^{r}\right] + \mathbb{E}\left[\varepsilon^{\frac{1}{\alpha}} \int_{\tau/\varepsilon}^{(\tau+\delta)/\varepsilon} \left|V_{u}\right|^{\gamma} u^{-\beta} \, \mathrm{d}u\right]^{r}.$$

The stopping time has a finite range  $\mathcal{T}$ , hence, we can write

$$\begin{split} \mathbb{E}\left[\left\|\int_{\tau/\varepsilon}^{(\tau+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_u \right\|^r\right] &= \mathbb{E}\left[\mathbb{E}\left[\left\|\int_{\tau/\varepsilon}^{(\tau+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_u \right\|^r \Big| \tau\right]\right] \\ &= \mathbb{E}\left[\sum_{\tau_i \in \tau} \frac{1}{\mathbb{P}(\tau=\tau_i)} \mathbb{E}\left[\mathbbm{1}_{\{\tau=\tau_i\}} \left\|\int_{\tau_i/\varepsilon}^{(\tau_i+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_u \right\|^r\right] \mathbbm{1}_{\{\tau=\tau_i\}}\right] \\ &\leq \mathbb{E}\left[\sum_{\tau_i \in \tau} \frac{1}{\mathbb{P}(\tau=\tau_i)} \mathbb{E}\left[\left\|\int_{\tau_i/\varepsilon}^{(\tau_i+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_u \right\|^r\right] \mathbbm{1}_{\{\tau=\tau_i\}}\right]. \end{split}$$

Besides, we can use the self-similarity of S, Lemma 5.2 in [DF13], and the fact that  $e^{tA}$  is a rotation matrix for any  $t \in \mathbb{R}$ , to compute for each  $\tau_i \in \mathcal{T}$  and  $\delta$  small enough

$$\mathbb{E}\left[\left\|\int_{\tau_i/\varepsilon}^{(\tau_i+\delta)/\varepsilon} \varepsilon^{\frac{1}{\alpha}} e^{-uA} \Gamma \, \mathrm{d}S_u\right\|^r\right] = \mathbb{E}\left[\left\|\int_{\tau_i}^{\tau_i+\delta} e^{-\frac{u}{\varepsilon}A} \Gamma \, \mathrm{d}S_u\right\|^r\right] \le C_r \delta^{\frac{r}{\alpha}}.$$

Since  $\tau \in [m, M]$  a.s., the last term can be handled as in Section 5.2 in [GL21b] using moment estimates of V (see Proposition 3.3) to have

$$\mathbb{E}\left[\varepsilon^{\frac{1}{\alpha}} \int_{\tau/\varepsilon}^{(\tau+\delta)/\varepsilon} K |V_u|^{\gamma} u^{-\beta} du\right]^r \le C_{m,M} \varepsilon^{r\beta - r\frac{\gamma + \alpha - 1}{\alpha}}.$$

Since  $\eta > 0$  and by Markov's inequality, we obtain for  $\delta$  and  $\varepsilon$  small enough

$$\mathbb{P}\left(\left\|Y_{\tau+\delta}^{(\varepsilon)} - Y_{\tau}^{(\varepsilon)}\right\| \ge a\right) \le \frac{C_r \delta^{\frac{r}{\alpha}} + C_{m,M} \varepsilon^{r\beta - r\frac{\gamma + \alpha - 1}{\alpha}}}{a^r} \le \eta.$$

Moreover, by Markov's inequality and the moment estimates again, we deduce that for all  $t \in [m, M]$ ,

$$\lim_{a \to +\infty} \limsup_{\varepsilon \to 0} \mathbb{P}\left(\left\|Y_t^{(\varepsilon)}\right\| \geq a\right) \leq \lim_{a \to +\infty} \limsup_{\varepsilon \to 0} \frac{\mathbb{E}\left[\left\|Y_t^{(\varepsilon)}\right\|^r\right]}{a^r} \leq \lim_{a \to +\infty} \frac{Ct^{\frac{r}{\alpha}}}{a^r} = 0.$$

By Corollary and Theorem 16.8 p. 175 in [Bil99], this concludes the proof of the tightness on every compact interval of  $(0, +\infty)$ .

We will now prove the convergence of the f.d.d. of  $Y^{(\varepsilon)}$ . Thanks to the previous lemma, this will yield the weak convergence on every compact set (see Theorem 13.1 p. 139 in [Bil99]) in the super-critical and critical regimes. The convergence in distribution on the whole space  $\mathcal{D}$  will follow from Theorem 16.7 p. 174 in [Bil99], since all processes considered are càdlàg.

## 3.3.1 Convergence of the f.d.d. in the super-critical regime

Assume that  $\alpha q > 1$ . Recall that  $(Y_t^{(\varepsilon)})_{t \geq \varepsilon t_0} := (\varepsilon^{\frac{1}{\alpha}} Y_{t/\varepsilon})_{t \geq \varepsilon t_0}$ .

Proof of Theorem 1.2 (i).

STEP 1. We first prove the convergence of the f.d.d. of the Wiener-Lévy integral appearing in  $(SDE_Y)$ .

Recall that the local martingale  $M^{(\varepsilon)}$  was defined in (22).

STEP 1A. We begin with the convergence in distribution of  $M_{s,t}^{(\varepsilon)}:=M_t^{(\varepsilon)}-M_s^{(\varepsilon)}$ , for  $\varepsilon t_0\leq s\leq t$ . To this end, we study the characteristic function  $\phi_{s,t}^{(\varepsilon)}$  of  $M_{s,t}^{(\varepsilon)}$ . Let us recall that  $\psi$  denotes the characteristic exponent of L, and is given, for all  $\xi\in\mathbb{R}$ , by

$$\psi(\xi) = -a|\xi|^{\alpha}.$$

The characteristic function of the Wiener-Lévy integral can be computed as p. 105 in [Sat99], hence one has, for all  $\xi := (u, v) \in \mathbb{R}^2$ ,

$$\begin{split} \phi_{s,t}^{(\varepsilon)}(\xi) &= \mathbb{E}\left(\exp\left[-iu\varepsilon^{\frac{1}{\alpha}}\int_{s/\varepsilon}^{t/\varepsilon}\sin(y)\,\mathrm{d}L_y + iv\varepsilon^{\frac{1}{\alpha}}\int_{s/\varepsilon}^{t/\varepsilon}\cos(y)\,\mathrm{d}L_y\right]\right) \\ &= \mathbb{E}\left(\exp\left[i\varepsilon^{\frac{1}{\alpha}}\int_{s/\varepsilon}^{t/\varepsilon}(-u\sin(y) + v\cos(y))\,\mathrm{d}L_y\right]\right) \\ &= \exp\left(\int_{s/\varepsilon}^{t/\varepsilon}\psi\left(\varepsilon^{\frac{1}{\alpha}}[-u\sin(y) + v\cos(y)]\right)\mathrm{d}y\right) \\ &= \exp\left(-a\varepsilon\int_{s/\varepsilon}^{t/\varepsilon}|-u\sin(y) + v\cos(y)|^{\alpha}\,\mathrm{d}y\right). \end{split}$$

Using Lemma B.2, we deduce that  $\phi_{s,t}^{(\varepsilon)}(\xi)$  converges, as  $\varepsilon \to 0$ , to

$$\exp\left(-a(t-s)\frac{1}{2\pi}\int_0^{2\pi}\left|-u\sin(y)+v\cos(y)\right|^{\alpha}dy\right).$$

STEP 1B. We now compute explicitly the scale parameter of the stable limiting process. We denote by  $\lambda$  the uniform probability distribution on the circle  $\mathbb{S}^1$ . Thanks to a change of variable and the symmetry of  $\lambda$ , setting  $\omega := \frac{\xi}{\|\xi\|}$  for  $\xi = (u, v) \in \mathbb{R}^2 \setminus \{0\}$ , we have

$$\frac{1}{2\pi} \int_0^{2\pi} \left| -u \sin(y) + v \cos(y) \right|^{\alpha} dy = \int_{\mathbb{S}^1} \left| \xi \cdot \lambda \right|^{\alpha} d\lambda$$
$$= \left\| \xi \right\|^{\alpha} \int_{\mathbb{S}^1} \left| \omega \cdot \lambda \right|^{\alpha} d\lambda.$$

Since  $\lambda$  is rotationally invariant, we deduce that  $\int_{\mathbb{S}^1} |\omega \cdot \lambda|^{\alpha} d\lambda$  does not depend on  $\omega \in \mathbb{S}^1$ . Taking  $\omega = (1,0)^T$ , we set

$$\widetilde{C} := \frac{a}{2\pi} \int_0^{2\pi} \left| \cos(x) \right|^{\alpha} \mathrm{d}x. \tag{23}$$

We have thus proved that, for any  $\xi \in \mathbb{R}^2$ ,

$$\phi_{s,t}^{(\varepsilon)}(\xi) \longrightarrow_{\varepsilon \to 0} \exp\left(-(t-s)\widetilde{C}\|\xi\|^{\alpha}\right).$$

Thus, the following convergence in distribution holds

$$M_{s,t}^{(\varepsilon)} = M_t^{(\varepsilon)} - M_s^{(\varepsilon)} \quad \Longrightarrow_{\varepsilon \to 0} \quad \mathcal{L}_{t-s}.$$
 (24)

Following the same lines, we show that, for any t > 0,

$$M_t^{(\varepsilon)} \underset{\varepsilon \to 0}{\Longrightarrow} \mathcal{L}_t.$$
 (25)

STEP 1C. We now prove the convergence in f.d.d. of  $M^{(\varepsilon)}$  to  $\mathcal{L}$ , as  $\varepsilon$  tends to 0.

Let us fix  $0 < t_1 \le t_2 \le \cdots \le t_d$ . Note that  $(M_t^{(\varepsilon)})_{t \ge \varepsilon t_0}$  is a càdlàg process with independent increments, since the integrands in its definition are deterministic and because L is a Lévy process. Thus, the random variables  $(M_{t_1}^{(\varepsilon)}, M_{t_1, t_2}^{(\varepsilon)}, \dots, M_{t_{d-1}, t_d}^{(\varepsilon)})$  are mutually independent. We deduce from the convergence results established in (24) and (25), and the fact that  $\mathcal{L}$  has stationary and independent increments that

$$(M_{t_1}^{(\varepsilon)}, M_{t_1, t_2}^{(\varepsilon)}, \dots, M_{t_{d-1}, t_d}^{(\varepsilon)}) \quad \Longrightarrow_{\varepsilon \to 0} \quad (\mathcal{L}_{t_1}, \mathcal{L}_{t_2} - \mathcal{L}_{t_1}, \dots, \mathcal{L}_{t_d} - \mathcal{L}_{t_{d-1}}).$$

The continuous mapping theorem yields the convergence in f.d.d. of  $M^{(\varepsilon)}$  to  $\mathcal{L}$ .

Step 2. Pick T > 0. We prove that

$$\mathbb{E}\left[\sup_{\varepsilon t_0 \leq t \leq T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \right] \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad 0.$$

We have

$$\sup_{\varepsilon t_0 \le t \le T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \le \varepsilon^{\frac{1}{\alpha}} \left\| z_0 \right\| + \varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{T/\varepsilon} \left\| e^{-sA} F(s, V_s) \right\| \mathrm{d}s.$$

We use moment estimates (Proposition 3.3) to get

$$\mathbb{E}\left[\varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{T/\varepsilon} \left\| e^{-sA} F(s, V_s) \right\| \mathrm{d}s \right] = \mathbb{E}\left[\varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{T/\varepsilon} \left\| F(s, V_s) \right\| \mathrm{d}s \right]$$

$$\leq \mathbb{E}\left[\varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{T/\varepsilon} \left| V_s \right|^{\gamma} s^{-\beta} \, \mathrm{d}s \right]$$

$$\leq \varepsilon^{\frac{1}{\alpha}} C_{\kappa, t_0} \int_{t_0}^{T/\varepsilon} s^{\frac{\gamma}{\alpha} - \beta} \, \mathrm{d}s$$

$$\leq C_{\kappa, t_0} \left(\varepsilon^{\beta - \frac{\gamma + \alpha - 1}{\alpha}} T^{\frac{\gamma}{\alpha} - \beta + 1} - \varepsilon^{\frac{1}{\alpha}} t_0^{\frac{\gamma}{\alpha} - \beta + 1} \right).$$

Hence, setting  $r := \min(\beta - \frac{\gamma + \alpha - 1}{\alpha}, \frac{1}{\alpha})$ , which is positive by assumption, we get

$$\mathbb{E}\left[\sup_{\varepsilon t_0 < t < T} \left\| Y_t^{(\varepsilon)} - M_t^{(\varepsilon)} \right\| \right] = \underset{\varepsilon \to 0}{O} (\varepsilon^r). \tag{26}$$

The conclusion follows from Theorem 3.1 p. 27 in [Bil99].

#### 3.3.2 Convergence of the f.d.d. in the critical and sub-critical regime

In this section, we consider the linear case, i.e.  $\gamma = 1$  and we assume that  $\beta \in (\frac{1}{2}, 1]$ . Recall that

$$(Y_t^{(\varepsilon)})_{t \ge \varepsilon t_0} = (\varepsilon^q Y_{t/\varepsilon})_{t > \varepsilon t_0}.$$

*Proof of Theorem 1.2 (ii) and (iii).* The proof follows the same lines as in the Brownian setting. Leaving out the noise, recall that the underlying ODE is the following

$$x''(t) + \frac{x'(t)}{t^{\beta}} + x(t) = 0, \quad t \ge t_0.$$
 (27)

We pick again the basis of solutions given by Lemma A.3, and we still denote by R its resolvent matrix and by f its rate of decrease. Recall that it is given by

$$\forall t > 0, f(t) := \begin{cases} \frac{1}{\sqrt{t}} & \text{if } \beta = 1, \\ \exp\left(-\frac{t^{1-\beta}}{2(1-\beta)}\right) & \text{else.} \end{cases}$$
 (28)

We set, for all  $t \geq \varepsilon t_0$ ,

$$\widetilde{M}_{t}^{(\varepsilon)} := \varepsilon^{q} f(t/\varepsilon) \int_{t_{0}}^{t/\varepsilon} R_{s}^{-1} \Gamma \, \mathrm{d}S_{s}. \tag{29}$$

Keeping the same notations as in the Brownian case, we decompose  $(Y_t)_{t\geq t_0}=(e^{-tA}Z_t)_{t\geq t_0}$  into

$$\varepsilon^q Y_{t/\varepsilon} = \varepsilon^q f(t/\varepsilon) \Phi_{t/\varepsilon} R_{t_0}^{-1} Z_0 + \Phi_{t/\varepsilon} \widetilde{M}_t^{(\varepsilon)}.$$

Reasoning as in the Brownian case, it remains to study the convergence of  $\widetilde{M}^{(\varepsilon)}$  since the first term converges towards 0. Using the expression of the Wronskian obtained in Lemma A.3, we obtain, for all  $t \geq \varepsilon t_0$ ,

$$\widetilde{M}_t^{(\varepsilon)} = \varepsilon^q f(t/\varepsilon) \int_{t_0}^{t/\varepsilon} f(u)^{-2} \begin{pmatrix} -y_2(u) \\ y_1(u) \end{pmatrix} dL_u.$$

Let us fix 0 < s < t. We study the convergence in distribution of the couple  $(\widetilde{M}_s^{(\varepsilon)}, \widetilde{M}_t^{(\varepsilon)})$  when  $\varepsilon$  tends to 0. The convergence in distribution of a general d-dimensional distribution  $(\widetilde{M}_{t_1}^{(\varepsilon)}, \ldots, \widetilde{M}_{t_d}^{(\varepsilon)})$  relies on the same computations. Let us fix  $(\xi_1, \xi_2) \in \mathbb{R}^2 \times \mathbb{R}^2$ . Using that L has independent increments, the characteristic function

Let us fix  $(\xi_1, \xi_2) \in \mathbb{R}^2 \times \mathbb{R}^2$ . Using that L has independent increments, the characteristic function  $\widetilde{\phi}_{s,t}^{(\varepsilon)}$  of  $(\widetilde{M}_s^{(\varepsilon)}, \widetilde{M}_t^{(\varepsilon)})$  is given by

$$\begin{split} \widetilde{\phi}_{s,t}^{(\varepsilon)}(\xi_1,\xi_2) &= \mathbb{E}\left[\exp\left(i\varepsilon^q\left[f\left(s/\varepsilon\right)\xi_1\cdot\int_{t_0}^{s/\varepsilon}f(u)^{-2}\begin{pmatrix}-y_2(u)\\y_1(u)\end{pmatrix}\mathrm{d}L_u\right.\right.\right.\\ &\left.\left.\left.\left.\left.\left.\left.\left.\left.\left.\left.f\left(t/\varepsilon\right)\xi_2\cdot\int_{t_0}^{t/\varepsilon}f(u)^{-2}\begin{pmatrix}-y_2(u)\\y_1(u)\end{pmatrix}\mathrm{d}L_u\right]\right)\right]\right]\right.\\ &= \mathbb{E}\left[\exp\left(i\varepsilon^q\left(f\left(s/\varepsilon\right)\xi_1+f\left(t/\varepsilon\right)\xi_2\right)\cdot\int_{t_0}^{s/\varepsilon}f(u)^{-2}\begin{pmatrix}-y_2(u)\\y_1(u)\end{pmatrix}\mathrm{d}L_u\right)\right]\right.\\ &\times \mathbb{E}\left[\exp\left(i\varepsilon^qf\left(t/\varepsilon\right)\xi_2\cdot\int_{s/\varepsilon}^{t/\varepsilon}f(u)^{-2}\begin{pmatrix}-y_2(u)\\y_1(u)\end{pmatrix}\mathrm{d}L_u\right)\right]. \end{split}$$

Let us recall that the characteristic exponent of L is given, for all  $\xi \in \mathbb{R}$ , by

$$\psi(\xi) = -a|\xi|^{\alpha}.$$

The characteristic function of the Wiener-Lévy integral can be computed as p. 105 in [Sat99], hence one has

$$\widetilde{\phi}_{s,t}^{(\varepsilon)}(\xi_1, \xi_2) = \exp\left(-a\varepsilon^{\beta} \int_{t_0}^{s/\varepsilon} f(u)^{-2\alpha} \left| \left(f\left(s/\varepsilon\right)\xi_1 + f\left(t/\varepsilon\right)\xi_2\right) \cdot \begin{pmatrix} -y_2(u) \\ y_1(u) \end{pmatrix} \right|^{\alpha} du \right) \times \exp\left(-a\varepsilon^{\beta} \int_{s/\varepsilon}^{t/\varepsilon} f(u)^{-2\alpha} \left| f\left(t/\varepsilon\right)\xi_2 \cdot \begin{pmatrix} -y_2(u) \\ y_1(u) \end{pmatrix} \right|^{\alpha} du \right).$$

Using the asymptotic expansion of the resolvent matrix (Lemma A.3), we can write, for any  $u \ge t_0$ ,

$$\begin{pmatrix} -y_2(u) \\ y_1(u) \end{pmatrix} = f(u) \left[ \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} + g(u) \right],$$

where  $g:[t_0,+\infty)\to\mathbb{R}^2$  is a function satisfying for all  $u\geq t_0$ ,

$$|g(u)| \le Cu^{1-2\beta}.$$

We set

$$K_1^{(\varepsilon)} := \exp\left(-a\varepsilon^{\beta} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} \left| (f(s/\varepsilon)\xi_1 + f(t/\varepsilon)\xi_2) \cdot \left[ \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} + g(u) \right] \right|^{\alpha} du \right)$$
(30)

and

$$K_2^{(\varepsilon)} := \exp\left(-a\varepsilon^{\beta} \int_{s/\varepsilon}^{t/\varepsilon} f(u)^{-\alpha} \left| f\left(\frac{t}{\varepsilon}\right) \xi_2 \cdot \left[ \left(-\sin(u) \cos(u)\right) + g(u) \right] \right|^{\alpha} du \right).$$

We thus obtain

$$\widetilde{\phi}_{s,t}^{(\varepsilon)}(\xi_1, \xi_2) = K_1^{(\varepsilon)} \times K_2^{(\varepsilon)}, \tag{31}$$

STEP 1. We start by justifying that we can omit g to study the limit when  $\varepsilon \to 0$ . More precisely, we prove that, for all function  $\zeta : \mathbb{R} \to \mathbb{R}^2$  such that  $\|\zeta(\varepsilon)\| f(s/\varepsilon)^{-1} = O(1)$ ,

$$R^{(\varepsilon)} := \varepsilon^{\beta} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} \left\| \left| \zeta(\varepsilon) \cdot \left[ \left( -\sin(u) \right) + g(u) \right] \right|^{\alpha} - \left| \zeta(\varepsilon) \cdot \left[ \left( -\sin(u) \right) \right] \right|^{\alpha} du \underset{\varepsilon \to 0}{\longrightarrow} 0. \quad (32)$$

Thanks to the mean value theorem applied to  $|\cdot|^{\alpha}$  (for  $\alpha \geq 1$ ) or by Hölder property (for  $\alpha < 1$ ), and the domination of g, we obtain that, for some constant C > 0,

$$R^{(\varepsilon)} \le C\varepsilon^{\beta} \|\zeta(\varepsilon)\|^{\alpha} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} u^{1-2\beta} du = \mathop{O}_{\varepsilon \to 0} (\|\zeta(\varepsilon)\|^{\alpha} f(s/\varepsilon)^{-\alpha} \varepsilon^{2\beta-1}),$$

where the last equality follows from Lemma B.3. This proves (32) since  $\beta > \frac{1}{2}$ .

STEP 2. We focus on the first term  $K_1^{(\varepsilon)}$  defined in (30). Since f is decreasing, notice that

$$\zeta(\varepsilon) := f(s/\varepsilon) \xi_1 + f(t/\varepsilon) \xi_2 = \mathop{O}_{\varepsilon \to 0} (f(s/\varepsilon)).$$

Then we have to study the convergence of  $I^{(\varepsilon)}$  defined by

$$I^{(\varepsilon)} := a\varepsilon^{\beta} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} \left| \left( f(s/\varepsilon) \xi_1 + f(t/\varepsilon) \xi_2 \right) \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du.$$

Its limit differs according to the value of  $\beta$ .

STEP 2A. Assume first that  $\beta = 1$ . Then, using the expression of f (see (28)),

$$I^{(\varepsilon)} = a\varepsilon^{1+\frac{\alpha}{2}} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} \left| \left( \frac{\xi_1}{\sqrt{s}} + \frac{\xi_2}{\sqrt{t}} \right) \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du.$$

We proved in Step 2b of the super-critical regime that there exists a constant  $\widetilde{C} > 0$  given in (23) such that, for all  $\zeta \in \mathbb{R}^2$ ,

$$\frac{a}{2\pi} \int_0^{2\pi} \left| \zeta \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du = \widetilde{C} \|\zeta\|^{\alpha}. \tag{33}$$

Using Lemma B.2, we can compute the following asymptotic expansion

$$I^{(\varepsilon)} = \varepsilon^{1+\frac{\alpha}{2}} \widetilde{C} \left\| \frac{\xi_1}{\sqrt{s}} + \frac{\xi_2}{\sqrt{t}} \right\|^{\alpha} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} du + \underset{\varepsilon \to 0}{o} \left( \varepsilon^{1+\frac{\alpha}{2}} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} du \right).$$

Therefore, it follows from Lemma B.3 that

$$K_1^{(\varepsilon)} \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad \exp\left(-\widetilde{C}\left(1+\frac{\alpha}{2}\right)^{-1} \left\|\frac{\xi_1}{\sqrt{s}} + \frac{\xi_2}{\sqrt{t}}\right\|^{\alpha} s^{1+\frac{\alpha}{2}}\right).$$

STEP 2B. Let us consider now  $\beta \in (\frac{1}{2}, 1)$ . Let us notice that  $I^{(\varepsilon)}$  can be decomposed into the sum

$$I^{(\varepsilon)} = I_1^{(\varepsilon)} + I_2^{(\varepsilon)} \tag{34}$$

of the two following terms

$$I_1^{(\varepsilon)} := a\varepsilon^{\beta} \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} \left| f\left(\frac{s}{\varepsilon}\right) \xi_1 \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du$$

and

$$I_{2}^{(\varepsilon)} := a\varepsilon^{\beta} \int_{t_{0}}^{s/\varepsilon} f(u)^{-\alpha} \left[ \left| \left( f\left( s/\varepsilon \right) \xi_{1} + f\left( t/\varepsilon \right) \xi_{2} \right) \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} - \left| f\left( s/\varepsilon \right) \xi_{1} \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} \right] du.$$

Using again either the mean value theorem or the Hölder property, and Lemma B.3, we get since  $\beta < 1$  that for some constant C > 0,

$$|I_2^{(\varepsilon)}| \le C\varepsilon^{\beta} f(s/\varepsilon)^{\alpha-1} f(t/\varepsilon) \int_{t_0}^{s/\varepsilon} f(u)^{-\alpha} du = \mathop{O}_{\varepsilon \to 0} \left( f(t/\varepsilon) f(s/\varepsilon)^{-1} \right) = \mathop{O}_{\varepsilon \to 0} (1). \tag{35}$$

Using Lemma B.2, we can compute the following asymptotic expansion of  $I_1^{(\varepsilon)}$ 

$$I_{1}^{(\varepsilon)} = a\varepsilon^{\beta} f\left(\frac{s}{\varepsilon}\right)^{\alpha} \int_{t_{0}}^{s/\varepsilon} f(u)^{-\alpha} \left| \xi_{1} \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du$$

$$= a\varepsilon^{\beta} f\left(\frac{s}{\varepsilon}\right)^{\alpha} \left[ \left(\frac{1}{2\pi} \int_{0}^{2\pi} \left| \xi_{1} \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du \right) \int_{t_{0}}^{s/\varepsilon} f(u)^{-\alpha} du + \underset{\varepsilon \to 0}{o} \left( \int_{t_{0}}^{s/\varepsilon} f(u)^{-\alpha} du \right) \right].$$

Thanks to (33) and the asymptotic expansion's results given in Lemma B.3, there exists an explicit constant  $k_{\beta,\alpha}$  given in Lemma B.3, such that

$$I_1^{(\varepsilon)} \xrightarrow[\varepsilon \to 0]{} k_{\beta,\alpha} \widetilde{C} s^{\beta} \|\xi_1\|^{\alpha}.$$
 (36)

Combining (32), (34), (35) and (36), we have proved that  $K_1^{(\varepsilon)}$ , defined in (30), converges as  $\varepsilon \to 0$  towards

$$\exp\left(-k_{\beta,\alpha}\widetilde{C}s^{\beta}\|\xi_1\|^{\alpha}\right).$$

Step 3. It remains to deal with the limit of  $K_2^{(\varepsilon)}$ . Notice that

$$\zeta(\varepsilon) := f(t/\varepsilon) \xi_2 = O(f(t/\varepsilon)) = O(f(s/\varepsilon)).$$

Hence, thanks to Step 1, we are reduced to study, for  $r \in \{s, t\}$ ,

$$J_r^{(\varepsilon)} := a\varepsilon^{\beta} f\left(\frac{t}{\varepsilon}\right)^{\alpha} \int_{t_0}^{r/\varepsilon} f(u)^{-\alpha} \left| \xi_2 \cdot \begin{pmatrix} -\sin(u) \\ \cos(u) \end{pmatrix} \right|^{\alpha} du. \tag{37}$$

Asymptotic expansion's results (Lemmas B.2 and B.3) and (33) yield

$$J_r^{(\varepsilon)} = \widetilde{C} \|\xi_2\|^{\alpha} \varepsilon^{\beta} f(t/\varepsilon)^{\alpha} \int_{t_0}^{r/\varepsilon} f(u)^{-\alpha} du + \underset{\varepsilon \to 0}{o} \left( f(t/\varepsilon)^{\alpha} f(r/\varepsilon)^{-\alpha} \right)$$
$$= \widetilde{C} \|\xi_2\|^{\alpha} k_{\beta,\alpha} r^{\beta} f(t/\varepsilon)^{\alpha} f(r/\varepsilon)^{-\alpha} + \underset{\varepsilon \to 0}{o} \left( f(t/\varepsilon)^{\alpha} f(r/\varepsilon)^{-\alpha} \right).$$

Hence,

$$J_{t}^{(\varepsilon)} \xrightarrow[\varepsilon \to 0]{} \widetilde{C} \|\xi_{2}\|^{\alpha} k_{\beta,\alpha} t^{\beta}$$
(38)

and

$$J_s^{(\varepsilon)} \xrightarrow[\varepsilon \to 0]{} \widetilde{C} \|\xi_2\|^{\alpha} k_{\beta,\alpha} s^{\beta} \left(\frac{s}{t}\right)^{\frac{\alpha}{2}} \mathbb{1}_{\{\beta=1\}}.$$
 (39)

Since

$$K_2^{(\varepsilon)} = \exp\left(-J_t^{(\varepsilon)} + J_s^{(\varepsilon)}\right),$$

we thus obtain that, for all  $0 < s \le t$ ,

$$\widetilde{\phi}_{s,t}^{(\varepsilon)}(\xi_1,\xi_2) \quad \underset{\varepsilon \to 0}{\longrightarrow} \quad \begin{cases} \exp\left(-k_{\beta,\alpha}\widetilde{C}s^{\beta}\|\xi_1\|^{\alpha}\right) \exp\left(-k_{\beta,\alpha}\widetilde{C}t^{\beta}\|\xi_2\|^{\alpha}\right) & \text{if } \beta < 1, \\ \exp\left(-k_{\beta,\alpha}\widetilde{C}\left[\left\|\frac{\xi_1}{\sqrt{s}} + \frac{\xi_2}{\sqrt{t}}\right\|^{\alpha}s^{1+\frac{\alpha}{2}} + \|\xi_2\|^{\alpha}t - \|\xi_2\|^{\alpha}\left(\frac{s}{t}\right)^{\frac{\alpha}{2}}s\right]\right) & \text{if } \beta = 1. \end{cases}$$

STEP 4. We can compute the characteristic function of the process  $\left(\frac{1}{\sqrt{t}}\int_0^t \sqrt{s} \, \mathrm{d}\mathcal{L}_s\right)_{t>0}$  in the same manner, and thus recognize the limiting process in the critical regime.

**Remark 3.6.** • As in the Brownian setting, if  $\beta = 0$ , the resolvent matrix is explicit and following the same lines, we can prove that  $(Z_{t/\varepsilon})_{t \geq \varepsilon t_0}$  converges in f.d.d. towards the product of the measure  $\mu$ , whose characteristic function is given by

$$\xi \mapsto \exp\left(-a\int_0^{+\infty} e^{-\alpha u} \|\xi \cdot h(u)\|^{\alpha} du\right),$$

h being an explicit periodic function depending on the resolvent matrix.

• As in the Brownian setting, one can prove the result in the linear case, i.e.  $\gamma = 1$ , following the same lines.

### 3.4 Proof of Corollary 1.5

Proof of Corollary 1.5. We start by proving the convergence in distribution of  $v_{1/T}Z_T$ . Reasoning as in the Brownian setting, it follows from Theorem 1.2 that  $v_{1/T}Y_T$  converges. The conclusion is a consequence of Lemma B.4, noting that the limiting distribution is invariant under rotations thanks to the expression of its characteristic function.

Let us now prove that the rescaled process  $Z^{(\varepsilon)}$  does not converge in distribution. We state the proof in the super-critical regime. Assume by contradiction that it is the case. Reasoning as in the Brownian case, we prove that this implies the convergence in distribution of the process  $I^{(\varepsilon)}$  defined, for  $t \geq \varepsilon t_0$ , by

$$I_t^{(\varepsilon)} := \varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{t/\varepsilon} \sin\left(\frac{t}{\varepsilon} - u\right) dL_u.$$

In particular, for s < t, the random variable  $I_t^{(\varepsilon)} - I_s^{(\varepsilon)}$  shall converge in distribution.

Let us denote by  $\phi^{(\varepsilon)}$  the characteristic function of  $I_t^{(\varepsilon)} - I_s^{(\varepsilon)}$ , which is supposed to converge on  $\mathbb{R}$ . Using that L has independent increments, we have

$$\phi^{(\varepsilon)}(1) = \mathbb{E}\left[\exp\left(\varepsilon^{\frac{1}{\alpha}} \int_{s/\varepsilon}^{t/\varepsilon} \sin\left(\frac{t}{\varepsilon} - u\right) dL_u\right)\right] \mathbb{E}\left[\exp\left(\varepsilon^{\frac{1}{\alpha}} \int_{t_0}^{s/\varepsilon} \sin\left(\frac{t}{\varepsilon} - u\right) - \sin\left(\frac{s}{\varepsilon} - u\right) dL_u\right)\right]$$

$$=: \phi^{(\varepsilon),1} \phi^{(\varepsilon),2}.$$

Recall that  $\psi$  defined in (2), denotes the characteristic exponent of L. Using a change of variables, we have in particular

$$\phi^{(\varepsilon),1} = \exp\left(\int_{s/\varepsilon}^{t/\varepsilon} \psi\left(\varepsilon^{\frac{1}{\alpha}} \sin\left(\frac{t}{\varepsilon} - u\right)\right) du\right) = \exp\left(-a\varepsilon \int_{s/\varepsilon}^{t/\varepsilon} \left|\sin\left(\frac{t}{\varepsilon} - u\right)\right|^{\alpha} du\right)$$
$$= \exp\left(-a\varepsilon \int_{0}^{(t-s)/\varepsilon} \left|\sin\left(u\right)\right|^{\alpha} du\right).$$

Lemma B.2 ensures that  $\phi^{(\varepsilon),1}$  has a limit when  $\varepsilon$  converges to 0. Similarly, we obtain

$$\phi^{(\varepsilon),2} = \exp\left(\int_{t_0}^{s/\varepsilon} \psi\left(\varepsilon^{\frac{1}{\alpha}} \sin\left(\frac{t}{\varepsilon} - u\right) - \sin\left(\frac{s}{\varepsilon} - u\right)\right) du\right)$$

$$= \exp\left(-a\varepsilon \int_{t_0}^{s/\varepsilon} \left|\sin\left(\frac{t}{\varepsilon} - u\right) - \sin\left(\frac{s}{\varepsilon} - u\right)\right|^{\alpha} du\right)$$

$$= \exp\left(-a2^{\alpha} \left|\sin\left(\frac{t - s}{2\varepsilon}\right)\right|^{\alpha} \varepsilon \int_{t_0}^{s/\varepsilon} \left|\cos\left(\frac{t + s}{2\varepsilon} - u\right)\right|^{\alpha} du\right)$$

$$= \exp\left(-a2^{\alpha} \left|\sin\left(\frac{t - s}{2\varepsilon}\right)\right|^{\alpha} \varepsilon \int_{\frac{t - s}{2\varepsilon}}^{\frac{t + s}{2\varepsilon} - t_0} \left|\cos(u)\right|^{\alpha} du\right). \tag{40}$$

The change of variables  $u = v + \pi$  yields, for all  $\varepsilon > 0$ ,

$$\int_0^{2\pi} |\cos(u)|^{\alpha} \operatorname{sgn}\left(\sin\left(\frac{t-s}{2\varepsilon}\right)\cos(u)\right) du = 0.$$

Thus, Lemma B.2 ensures that

$$\varepsilon \int_{\frac{t-s}{2\varepsilon}}^{\frac{t+s}{2\varepsilon}-t_0} |\cos(u)|^{\alpha} du \quad \xrightarrow{\varepsilon \to 0} \quad \frac{s}{2\pi} \int_{0}^{2\pi} |\cos(u)|^{\alpha} du.$$

Coming back to (40), we see that  $\phi^{(\varepsilon),2}$  does not converge when  $\varepsilon$  tends to 0. This is a contradiction.

## A Study of the deterministic underlying ODE

The deterministic ODE behind the system, i.e. without frictional force and without noise, is the following

$$x''(t) + \frac{x'(t)}{t^{\beta}} + x(t) = 0, \quad t \ge t_0.$$
(41)

The solutions form a vector space of dimension 2. Let us take two solutions  $y_1$  and  $y_2$  which are linearly independent. Then, we introduce the fundamental system of solutions (resolvent matrix) R to (41) defined, for  $t \geq t_0$ , by

$$R_t = \begin{pmatrix} y_1(t) & y_2(t) \\ y_1'(t) & y_2'(t) \end{pmatrix}.$$

It satisfies, for all  $t \geq t_0$ ,

$$R_t' = \begin{pmatrix} 0 & 1 \\ -1 & -\frac{1}{t^{\beta}} \end{pmatrix} R_t.$$

We recall that the Wronskian w is defined, for all  $t \geq t_0$ , by

$$w(t) = y_1(t)y_2'(t) - y_1'(t)y_2(t).$$

Let us finally set, for t > 0,

$$f(t) := \begin{cases} \frac{1}{\sqrt{t}} & \text{if } \beta = 1, \\ \exp\left(-\frac{t^{1-\beta}}{2(1-\beta)}\right) & \text{else.} \end{cases}$$
 (42)

**Lemma A.1.** Pick  $\beta \in (\frac{1}{2}, +\infty)$  and consider a solution y to (41). Then, there exist  $a \in \mathbb{R}$  and  $\phi \in [0, 2\pi)$  such that

$$y(t) = af(t)\cos(t+\phi) + O_{t\to\infty}\left(f(t)t^{-(2\beta-1)\wedge\beta}\right)$$

and

$$y'(t) = -af(t)\sin(t+\phi) + O_{t\to\infty}\left(f(t)t^{-(2\beta-1)\wedge\beta}\right).$$

*Proof.* Let us set, for  $t \ge t_0$ ,  $u(t) = f(t)^{-1}y(t)$ . We easily check that u satisfies

$$u''(t) + u(t) [1 + h(t)] = 0,$$

where  $h(t) := \frac{f''(t)}{f(t)} + \frac{f'(t)}{f(t)t^{\beta}} = \underset{t \to +\infty}{O}(t^{-(2\beta) \wedge (\beta+1)})$ . Following the proof of the method of variation of parameters, there exists  $a_0, b_0 \in \mathbb{R}$  such that, for any  $t \geq t_0$ ,

$$u(t) = a_0 \cos(t) + b_0 \sin(t) - \int_{t_0}^t u(s)h(s)\sin(t-s) ds.$$

Using that  $h \in L^1((t_0, +\infty))$  since  $\beta > \frac{1}{2}$ , we obtain by Grönwall's lemma that u is bounded on  $[t_0, +\infty)$ . We deduce that the functions  $s \mapsto u(s)h(s)\cos(s)$  and  $s \mapsto u(s)h(s)\sin(s)$  belong to  $L^1((t_0, +\infty))$ . Thus, up to changing the constants  $a_0$  and  $b_0$ , one has, for all  $t \geq t_0$ ,

$$u(t) = a_0 \cos(t) + b_0 \sin(t) - \sin(t) \int_t^\infty u(s)h(s)\cos(s) \, ds + \cos(t) \int_t^\infty u(s)h(s)\sin(s) \, ds.$$
 (43)

It follows from the fact that u is bounded that

$$u(t) = a_0 \cos(t) + b_0 \sin(t) + \mathop{O}_{t \to +\infty} \left( \int_t^\infty \frac{\mathrm{d}s}{s^{(2\beta) \land (\beta+1)}} \right).$$

Thus, there exist  $a \in \mathbb{R}$  and  $\phi \in [0, 2\pi)$  such that

$$u(t) = a\cos(t+\phi) + \mathop{O}_{t\to+\infty}(t^{-(2\beta-1)\wedge\beta}).$$

This proves the asymptotic expansion of y. Differentiating (43) and using that  $h(t) = O(t^{-(2\beta) \wedge (\beta+1)})$ , we prove that

$$u'(t) = -a\sin(t+\phi) + O_{t\to +\infty}(t^{-(2\beta-1)\wedge\beta}).$$

Since u is bounded and  $f'(t) = \underset{t \to +\infty}{O} (f(t)t^{-\beta})$ , we finally obtain that

$$y'(t) = f'(t)u(t) + f(t)u'(t) = f(t)u'(t) + O_{t \to +\infty}(f(t)t^{-\beta}).$$

This concludes the proof of the asymptotic expansion of y'.

**Remark A.2.** Note that if  $\beta = 1$ , the Bessel functions of the first kind  $J_0$  and of the second kind  $Y_0$  form a basis of solutions. Their asymptotic expansions can be found in [Wat44, Chap VII].

**Lemma A.3.** Pick  $\beta \in (\frac{1}{2}, +\infty)$ . There exists a basis of solutions  $y_1$  and  $y_2$  to (41) such that the resolvent matrix R satisfies

$$R_t = f(t) \begin{pmatrix} \cos(t) & \sin(t) \\ -\sin(t) & \cos(t) \end{pmatrix} + \mathop{O}_{t \to \infty} \left( f(t) t^{-(2\beta - 1) \wedge \beta} \right) = f(t) e^{tA} + \mathop{O}_{t \to \infty} \left( f(t) t^{-(2\beta - 1) \wedge \beta} \right).$$

Moreover, its Wronskian w is given for any  $t \ge t_0$  by

$$w(t) = f(t)^2$$
.

*Proof.* It is well-known that the Wronskian satisfies, for all  $t \geq t_0$ ,

$$w'(t) = -\frac{1}{t^{\beta}}w(t).$$

Thus, there exists  $w_0 \in \mathbb{R} \setminus \{0\}$  such that, for all  $t \geq t_0$ ,  $w(t) = w_0 f(t)^2$ . Moreover, thanks to Lemma A.1, for  $i \in \{1, 2\}$ , there exist  $a_i \in \mathbb{R}$  and  $\phi_i \in [0, 2\pi)$  such that

$$y_i(t) = a_i f(t) \cos(t + \phi_i) + O_{t \to \infty} \left( f(t) t^{-(2\beta - 1) \land \beta} \right)$$

and

$$y_i'(t) = -a_i f(t) \sin(t + \phi_i) + O_{t \to \infty} \left( f(t) t^{-(2\beta - 1) \wedge \beta} \right).$$

As a consequence,

$$w(t) = -a_1 a_2 f(t)^2 \sin(\phi_2 - \phi_1) + O_{\text{total}} \left( f(t)^2 t^{-(4\beta - 2) \wedge 2\beta} \right).$$

But since  $w(t) = w_0 f(t)^2$ , it implies that  $a_i \neq 0$  and  $\phi_2 \not\equiv \phi_1[\pi]$ .

Up to dividing by  $a_i$ , we can assume that  $a_i = 1$ , and up to considering a linear combination of  $y_1$  and  $y_2$ , we can assume that  $\phi_1 = 0$  and  $\phi_2 = -\frac{\pi}{2}$ . Thus, we have  $w_0 = 1$ . This concludes the proof.

## B Some technical results

We collect here some technical results used in our proofs. Recall first a sufficient condition for the non-explosion of the solution to a SDE. The proof can be found in [GL21a].

**Lemma B.1.** Let  $(Y_t)_{t \geq t_0}$  be a càdlàg process, solution to a SDE. For all  $n \geq 0$ , define the stopping time

$$\tau_n := \inf\{t \ge t_0, \ \|Y_t\| \ge n\}. \tag{44}$$

Set  $\tau_{\infty} := \lim_{n \to +\infty} \tau_n$  the explosion time of Y. Assume that there exist two measurable and non-negative functions  $\phi$  and b such that

- (i)  $\phi$  is non-decreasing and  $\lim_{n\to\infty} \phi(n) = +\infty$ ,
- (ii) b is finite-valued,

(iii) and for all 
$$t \geq t_0$$
,

$$\sup_{n>0} \mathbb{E}\left[\phi(|Y_{t\wedge\tau_n}|)\right] \le b(t).$$

Then  $\tau_{\infty} = +\infty$  a.s.

We now state and prove a result on the periodic-averaging phenomenon.

**Lemma B.2.** Let us fix  $t_0 > 0$  and  $h : [t_0, +\infty) \to \mathbb{R}$  a continuous m-periodic function, with m > 0. Let  $g : [t_0, +\infty) \to \mathbb{R}^+$  be a continuously differentiable function which is not integrable on  $[t_0, +\infty)$ . We assume moreover that

(i) 
$$g'(t) = o_{t \to +\infty}(g(t)),$$

(ii) 
$$g(t) = \underset{t \to +\infty}{o} \left( \int_{t_0}^t g(u) \, \mathrm{d}u \right).$$

Then,

$$\int_{t_0}^t g(u)h(u)\,\mathrm{d}u = \left[\frac{1}{m}\int_{t_0}^{t_0+m}h(u)\,\mathrm{d}u\right]\int_{t_0}^t g(u)\,\mathrm{d}u + \mathop{o}_{t\to+\infty}\left(\int_{t_0}^t g(u)\,\mathrm{d}u\right).$$

Let us remark that the functions  $g_1$  and  $g_2$  defined for  $t \in \mathbb{R}$ ,  $r \geq 0$  and  $\beta \in [0,1)$  by  $g_1(t) := t^r$  and  $g_2(t) := \exp(rt^{1-\beta})$ , satisfy the preceding assumptions made on g.

*Proof.* Let us define  $h := h - \frac{1}{m} \int_{t_0}^{t_0+m} h(u) du$ , and H a primitive of h. The function H is bounded on  $[t_0, +\infty)$  since the average of h on its period is equal to 0. To prove the lemma, we only need to justify that

$$\int_{t_0}^t g(u)\widetilde{h}(u) du = \mathop{o}_{t \to +\infty} \left( \int_{t_0}^t g(u) du \right).$$

By integration by parts, we obtain that, for all  $t \geq t_0$ ,

$$\int_{t_0}^t g(u)\widetilde{h}(u) du = g(t)\widetilde{H}(t) - g(t_0)\widetilde{H}(t_0) - \int_{t_0}^t g'(u)\widetilde{H}(u) du.$$

Using the fact that  $\widetilde{H}$  is bounded, that  $g'(t) = \underset{t \to +\infty}{o}(g(t))$  and that  $\int_{t_0}^{\infty} g(u) du = +\infty$ , we deduce that

$$\int_{t_0}^t g'(u)\widetilde{H}(u) du = \mathop{o}_{t \to +\infty} \left( \int_{t_0}^t g(u) du \right).$$

The conclusion follows from the fact that  $g(t)\widetilde{H}(t)-g(t_0)\widetilde{H}(t_0)=\underset{t\to+\infty}{o}\left(\int_{t_0}^tg(u)\,\mathrm{d}u\right)$ , since we have assumed that  $g(t)=\underset{t\to+\infty}{o}\left(\int_{t_0}^tg(u)\,\mathrm{d}u\right)$  and that  $\int_{t_0}^\infty g(u)\,\mathrm{d}u=+\infty$ .

**Lemma B.3.** Let f be given by (42) for  $\beta \in (\frac{1}{2}, 1]$ , and pick  $\alpha \in (1, 2]$ . Define

$$k_{\beta,\alpha} := \begin{cases} (1+\alpha/2)^{-1} & \text{if } \beta = 1, \\ \frac{2}{\alpha} & \text{else.} \end{cases}$$

Then for any t > 0, we have

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{1-2\beta} \, \mathrm{d}u = \mathop{O}_{\varepsilon \to 0} (f(t/\varepsilon)^{-\alpha} \varepsilon^{\beta-1}),$$

and

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} du = k_{\beta,\alpha} f\left(\frac{t}{\varepsilon}\right)^{-\alpha} \left(\frac{t}{\varepsilon}\right)^{\beta} + \underset{\varepsilon \to 0}{o} (f(t/\varepsilon)^{-\alpha} \varepsilon^{-\beta}).$$

*Proof.* When  $\beta = 1$ , the results follow from direct computations because of the expression of f. Assume now that  $\beta \in (\frac{1}{2}, 1)$ . For the first point, the integration by parts formula ensures that

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{1-2\beta} du = \frac{2}{\alpha} \left[ f(u)^{-\alpha} u^{1-\beta} \right]_{t_0}^{t/\varepsilon} - \frac{2}{\alpha} (1-\beta) \int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{-\beta} du$$

$$= \underset{\varepsilon \to 0}{O} (f(t/\varepsilon)^{-\alpha} \varepsilon^{\beta-1}) + \underset{\varepsilon \to 0}{O} (f(t/\varepsilon)^{-\alpha})$$

$$= \underset{\varepsilon \to 0}{O} (f(t/\varepsilon)^{-\alpha} \varepsilon^{\beta-1}).$$

For the second asymptotic expansion, it follows again from an integration by parts that

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} du = \int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{-\beta} u^{\beta} du$$
$$= \frac{2}{\alpha} \left[ f(u)^{-\alpha} u^{\beta} \right]_{t_0}^{t/\varepsilon} - \frac{2}{\alpha} \beta \int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{\beta - 1} du.$$

Remarking that  $f(u)^{-\alpha}u^{\beta-1} = \underset{u \to +\infty}{o}(f(u)^{-\alpha})$ , since  $\beta < 1$ , we deduce that

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} u^{\beta-1} du = \underset{\varepsilon \to 0}{o} \left( \int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} du \right).$$

We obtain that

$$\int_{t_0}^{t/\varepsilon} f(u)^{-\alpha} du \underset{\varepsilon \to 0}{\sim} \frac{2}{\alpha} f\left(\frac{t}{\varepsilon}\right)^{-\alpha} \left(\frac{t}{\varepsilon}\right)^{\beta}$$

This ends the proof.

**Lemma B.4.** Let  $(X_n)_n$  be a sequence of random variables with values in  $\mathbb{R}^2$ , and which converges in distribution to a random variable X. We assume that the distribution of X is invariant under rotations, i.e. for any orthogonal matrix  $R \in \mathcal{M}_2(\mathbb{R})$ , the random variables X and RX have the same distribution. Then for all sequence  $(R_n)_n$  of orthogonal matrices in  $\mathcal{M}_2(\mathbb{R})$ , we have

$$R_n X_n \quad \Longrightarrow_{n \to +\infty} \quad X.$$

*Proof.* Let us denote by  $\phi_Z$  the characteristic function of a random variable Z. Using Theorem 5.3 p. 86 in [Kal02], we know that  $(\phi_{X_n})_n$  converges to  $\phi_X$  uniformly on every compact subset of  $\mathbb{R}^2$ . The characteristic function of the random variable  $Y_n := R_n X_n$  is given by

$$\xi \mapsto \phi_{Y_n}(\xi) = \phi_{X_n}(R_n \xi).$$

Thus, by assumption, we have, for all  $\xi \in \mathbb{R}^2$ ,

$$\phi_X(R_n\xi) = \phi_X(\xi).$$

It follows that, for any  $\xi \in \mathbb{R}^2$  and  $n \geq 0$ ,

$$|\phi_{Y_n}(\xi) - \phi(\xi)| = |\phi_{X_n}(R_n\xi) - \phi_X(R_n\xi)| \le \sup_{z \in \mathbb{R}^2, ||z|| = ||\xi||} |\phi_{X_n}(z) - \phi_X(z)|,$$

which converges to 0, as  $n \to +\infty$ . This ends the proof of the lemma.

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