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*Research article*

## Solvability and stability of mean-field stochastic differential equations driven by time-changed Lévy noise

Mahmoud Abouagwa<sup>1,2,\*</sup> and Maher Ibrahim Tawdrous<sup>3,4</sup>

<sup>1</sup> College of Business, City University Ajman, Ajman 18484, United Arab Emirates

<sup>2</sup> Department of Mathematical Statistics, Faculty of Graduate Studies for Statistical Research, Cairo University, Giza 12613, Egypt

<sup>3</sup> College of Humanities, City University Ajman, Ajman 18484, United Arab Emirates

<sup>4</sup> Faculty of Education, Suez Canal University, Ismailia 41522, Egypt

\* **Correspondence:** Email: mahmoud.aboagwa@cu.edu.eg, m.abouagwa@cu.ac.ae.

**Abstract:** In this article, we focused on a class of mean-field stochastic differential equations driven by time-changed Lévy noise. We first discussed the existence and uniqueness of solutions under the non-Lipschitz case with the Lipschitz condition as the special case by adopting the Carathéodory approximation. To prove our results, we established a new time-changed retarded integral inequality, which is easy to apply in practice and can be considered as a more general tool in some situations. Then, the classical Itô formula was extended to that for mean-field stochastic differential equations driven by time-changed Lévy noise. As an application of Itô's formula, we showed that the trivial solution is  $p$ -th moment asymptotically stable, stable in probability, asymptotically stable in probability, and globally asymptotically stable in probability based on the Lyapunov function. Finally, an example was presented to validate the produced results.

**Keywords:** mean-field stochastic differential equations; stability; time-changed Lévy noise; Itô's formula

**Mathematics Subject Classification:** 60B10, 60H10, 35Q83, 93E15

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### 1. Introduction

In light of the innovative research of Kac [13], McKean [26] considered the nonlinear Fokker-Plank equations using mean-field stochastic differential equations (MFSDEs), also named McKean-Vlasov or distribution dependent SDEs in the literature, in which the drift and diffusion coefficients depend not only on the state of the unknown process but also on its probability law (distribution). These equations arise as the limiting behavior of a representative particle from a mean-field interacting particle system

as the number of particles tends to infinity. For instance, we utilize the  $N$ -particles model described by SDEs

$$dY^{i,N}(t) = f(t, Y^{i,N}(t), \mathcal{L}^N(Y^{(N)}(t)))dt + g(t, Y^{i,N}(t), \mathcal{L}^N(Y^{(N)}(t)))dW_t^i, \quad (1.1)$$

where  $\mathcal{L}^N(Y^{(N)}(t)) := \frac{1}{N} \sum_{\ell=1}^N \delta_{Y^{\ell,N}(t)}$  refers to the empirical measure of the system at time  $t$ . The mean-field limit of (1.1) can be expected as

$$dY(t) = f(t, Y(t), \mathcal{L}(Y(t)))dt + g(t, Y(t), \mathcal{L}(Y(t)))dW_t$$

as  $N$  tends to infinity and  $\mathcal{L}^N(Y^{(N)}(t))$  converges weakly to  $\mathcal{L}(Y(t))$  with probability 1. Hence, the coefficients depend not only on the solution pointwisely but also on its time marginal law. This was named “propagation of chaos” by Kac [13].

Due to the investigation of complex networked systems [14], mean-field games [1, 2, 47], stochastic control [4], and statistical physics [15], there are many papers devoted to the study of mean-field systems and the relevant theories have been well-developed (see, for example, [34–36]). So far, researchers have shown an increased interest to the study of solvability and stability of MFSDs. For example, Bahlali et al. [4] studied the existence and uniqueness for solutions of MFSDs driven by Brownian motion under an Osgood-type condition. Also, they studied the stability with respect to initial data, coefficients, and driving processes. Qiao and Gong [37] established the existence, uniqueness, mean-square asymptotic stability, and almost surely asymptotic stability for solutions of non-Lipschitz multivalued MFSDs. Hammersley et al. [9] established the existence of weak solutions to mean-field SDEs under Lyapunov-type conditions. Huang [11] introduced the nonlinear Fokker-Planck equations for probability measures on path space and path-distribution dependent SDEs. Mezerdi [29] proposed the stability of mean-field SDEs under a non-Lipschitz condition. Mishura and Veretennikov [30] proved new weak and strong existence and uniqueness results for multi-dimensional mean-field SDEs driven by Brownian motion under relaxed regularity conditions. Ren and Wang [38] studied the path independence of Girsanov transformations of mean-field SDEs driven by Brownian motion. Shen et al. [41] studied the stochastic averaging principle and stability for multi-valued McKean-Vlasov SDEs with jumps.

In recent years, the theory of time-changed semimartingales has received increased attention and became a major area of interest within the field of stochastic analysis due to its wide range of applications in modeling anomalous diffusions appearing in physics, finance, cell biology, and hydrology [22, 42]. It arises as the scaling limit of continuous time random walks [27, 28]. Since 2011, the stochastic integrals with respect to time-changed semimartingales, the time-changed Itô formula, and the duality theorem between non-time-changed stochastic differential equations (SDEs), and the associated time-changed SDEs have been investigated by Kobayashi [16]. Since then, time-changed SDEs have been used to model some real-world phenomena such as Black-Scholes models. Particularly, the stocks that are not actively traded, where their prices stay constant for some period of time can be described by time-changed Brownian motion but not by the standard Brownian motion [33]. Through the past decade, considerable research has grown up around the theory of time-changed SDEs, and we would like to mention a few here. Wu [44] studied the exponential sample-path stability,  $p$ -th moment asymptotic stability, and  $p$ -th moment exponential stability for SDEs driven by time-changed Brownian motion. Nane and Ni [31] established the stabilities in both probability

and moment sense for SDEs driven by time-changed Lévy noise. For more work on SDEs driven by time-changed Brownian motion or time-changed Lévy noise, we refer the reader to [20, 32, 45, 46, 49].

Although extensive research has been carried out on the classical MFSDs, little is known on time-changed MFSDs. For example, Shen et al. [40] established the existence and uniqueness of strong solutions for MFSDs driven by the time-changed Brownian motion under the one-sided Lipschitz condition. They also gave sufficient conditions for the solution to be stochastically stable, stochastically asymptotically stable, and globally stochastically asymptotically stable using the Lyapunov function. Li et al. [21] considered the existence and stability with respect to the initial data and coefficients for impulsive and non-impulsive MFSDs driven by time-changed Brownian motion under the Lipschitz condition by means of the fixed point theorem. Wen et al. [43] provided the strong convergence on the Euler-Maruyama method for non-autonomous MFSDs driven by time-changed Brownian motion. Finally, Li and Ren [19] considered the practical stability with respect to part of the variables for MFSDs driven by time-changed Brownian motion.

On the other hand, in view of rapid changes in perturbation theory, the continuous sample paths of time-changed Brownian motion make it difficult to describe stochastic influences with possible jumps in financial economics, mathematical finance, stochastic filtering, and stochastic control, etc. Consequently, many real systems are very complex and may have discontinuous sample paths, which means that MFSDs perturbed with time-changed Brownian motion fail to describe them. It is noted that MFSDs perturbed with the time-changed Lévy noise are quite convenient to cope with such discontinuous systems [8]. Further, Lipschitz or one-sided Lipschitz assumptions are always considered as the assumptions to investigate the existence, uniqueness, and stability for solutions of MFSDs driven by time-changed Brownian motion (see [19, 21, 40]). However, these assumptions are no longer suitable to deal with various situations in the real world and many practical models of SDEs need some weaker conditions. For instance, the one-dimensional semi-linear SDEs with Markovian switching,

$$dY(t) = \tilde{f}(r(t))Y(t)dt + \tilde{g}(r(t))\tilde{h}(|Y(t)|)dW_t,$$

where  $W_t$  represents Brownian motion,  $r(t)$  is a continuous-time Markov chain, and  $\tilde{h} : \mathbb{R}^+ \rightarrow \mathbb{R}^+$ , is defined in the following manner:

$$\tilde{h}(v) = \begin{cases} v \sqrt{-\ln(v)}, & 0 \leq v \leq e^{-1}, \\ \delta \log(\delta^{-1}) + \kappa_2(\delta^-)(v - \delta)e^{-1} + \frac{1}{2}(v - e^{-1}), & v > e^{-1}. \end{cases}$$

Such models arise in various branches of science, engineering, industry, and finance [7, 10, 18]. Then, it is necessary to consider some weaker and more generalized conditions, such as the so-called non-Lipschitz condition [25, 29, 37].

Based on the above discussion, we will make the first attempt to consider the following class of non-Lipschitz MFSDs driven by time-changed Lévy noise:

$$\begin{aligned} dY(t) &= f(t, E_t, Y(t-), \mathcal{L}(Y(t-)))dE_t + g(t, E_t, Y(t-), \mathcal{L}(Y(t-)))dW_{E_t} \\ &+ \int_{|z|<c} \tilde{h}(t, E_t, Y(t-), \mathcal{L}(Y(t-)), z)\tilde{N}(dE_t, dz), \quad t \in J = [0, T], \end{aligned} \quad (1.2)$$

on a complete probability space  $(\Omega, \mathfrak{F}, \{\mathfrak{F}_t\}_{t \geq 0}, \mathbb{P})$  with a right continuous and increasing filtration  $\{\mathfrak{F}_t\}_{t \geq 0}$ , where  $\mathfrak{F}_0$  contains all  $\mathbb{P}$ -null sets. The coefficients  $f : J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}(\mathbb{R}^n) \mapsto \mathbb{R}^n$ ,

$g : J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}(\mathbb{R}^n) \mapsto \mathbb{R}^{n \times m}$ , and  $\tilde{h} : J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}(\mathbb{R}^n) \times (\mathbb{R}^n \setminus \{0\}) \mapsto \mathbb{R}^n$  are Borel measurable functions,  $\mathcal{L}(Y(t))$  stands for the probability law of the random variable  $Y(t)$ , and  $\mathcal{M}(\mathbb{R}^n)$  denotes the space of all probability measures on the Borel measurable space  $\mathbb{R}^n$  equipped with the weak topology. The initial value  $Y(0) = Y_0 \in \mathbb{R}^n$  is an  $\mathfrak{F}_0$ -measurable random variable satisfying  $\mathbb{E}|Y_0|^2 < \infty$ . The time-change  $E_t$  is the inverse of a  $\beta$ -stable subordinator and the composition  $(W_{E_t})_{t \in J}$  is called a time-changed Brownian motion, which is understood as a subdiffusion (see [27, 42]).  $\tilde{N}(dE_t, dz)$  is the compensated Poisson random measure. The rigorous mathematical definitions are postponed to Section 2.

The main innovative points of this article are listed below:

- Our model MFSDs (1.2) driven by time-changed Lévy noise are new to the Literature. Further, we introduce a new time-changed retarded integral inequality, which is easy to apply in practice and can be considered as a new contribution to the literature.
- The existence results in [21, 40] are obtained under Lipschitz and one-sided Lipschitz conditions. However, our existence and uniqueness results for Eq (1.2) are considered under a non-Lipschitz condition by using Carathéodory approximations with more complex computations, generalizing those of [21, 40].
- We extend the generalized Itô formula from classical SDEs to non-Lipschitz MFSDs driven by time-changed Lévy noise, generalizing and containing that of [40] as a special case.
- Since the stability analysis is of great interest in SDE theory, we present sufficient conditions for the solution of Eq (1.2) to be  $p$ -th moment asymptotically stable, stable in probability, asymptotically stable in probability, and globally asymptotically stable in probability.
- Moreover, letting  $\tilde{h} \equiv 0$  in Eq (1.2), the  $p$ -th moment asymptotic stability result in Theorem 5.1 is still new for MFSDs driven by time-changed Brownian motion and considered by [21, 40].
- Finally, an example is given to justify our Conditions 3.1–3.3 and stability results.

The rest of this article is organized as follows. In Section 2, we recall some necessary notations and preliminaries for this article. In Section 3, we devote ourselves to proving the existence and uniqueness theorem of solutions for our model (1.2) based on a new integral inequality given in Lemma 3.1 and Conditions 3.1–3.3. In Section 4, we state and prove the generalized Itô formula for non-Lipschitz MFSDs driven by time-changed Lévy noise. We present the stability results followed by an example to show the utility of the acquired theoretical results in Section 5.

## 2. Preliminaries

In this section, we briefly give some preliminaries that will be used in the following sections. Let  $(\Omega, \mathfrak{F}, \{\mathfrak{F}_t\}_{t \geq 0}, \mathbb{P})$  be a complete probability space with a filtration  $\{\mathfrak{F}_t\}_{t \geq 0}$  satisfying the usual conditions (i.e., it is increasing and right continuous, while  $\mathfrak{F}_0$  contains all  $\mathbb{P}$ -null sets). Assume  $W = (W_t)_{t \geq 0}$  represents an  $m$ -dimensional  $\{\mathfrak{F}_t\}_{t \geq 0}$ -Brownian motion and  $\{\mathcal{D}(t), t \geq 0\}$  is a right continuous with a left-limit-increasing  $\{\mathfrak{F}_t\}_{t \geq 0}$ -Lévy process with a Lévy symbol  $1 < \beta < 2$ , called a  $\beta$ -stable subordinator, starting from  $\mathcal{D}(0) = 0$  with the Laplace transform  $\mathbb{E}(e^{-\lambda \mathcal{D}(\tau)}) = e^{-\tau \phi(\lambda)}$ ,  $\lambda > 0$ , where the Laplace exponent  $\phi(\lambda) = \int_0^\infty (1 - e^{-\lambda y}) \alpha(dy)$  is associated with a given  $\sigma$ -finite measure  $\alpha$  on  $(0, \infty)$  such that  $\int_0^\infty (1 \wedge y) \alpha(dy) < \infty$ . Suppose that  $\alpha$  is infinite, i.e.,  $\alpha(0, \infty) = \infty$ . The stochastic time-change  $E_t = \inf\{\tau > 0 : \mathcal{D}(\tau) > t\}$  is independent of  $W$  and can be defined as the inverse subordinator of  $\mathcal{D}(t)$ .

Moreover,  $E_t$  is continuous and non-decreasing, however, it is not Markovian and can be considered as the first hitting time process. Suppose that  $W_t$  is independent of  $\mathcal{D}(t)$ . Without loss of generality, we clarify that the filtration  $\{\mathfrak{F}_t\}_{t \geq 0}$  is determined by

$$\mathfrak{F}_t = \bigcap_{\tau > t} \left\{ \sigma_1[W_s : 0 \leq s \leq \tau] \vee \sigma_2[E_s : s \geq 0] \right\},$$

where the notation  $\sigma_1 \vee \sigma_2$  refers to the  $\sigma$ -algebra generated by the union of the two  $\sigma$ -algebras  $\sigma_1$  and  $\sigma_2$ .

The composition  $W \circ E = (W_{E_t})_{t \geq 0}$ , called a time-changed Brownian motion [23], is a square integrable martingale with respect to the natural filtration  $\{\mathcal{G}_t\}_{t \geq 0}$ , where  $\mathcal{G}_t := \mathfrak{F}_{E_t}$ ,  $t \geq 0$ , for the process  $\{E_t\}$ . Let  $N$  be the  $\mathfrak{F}_t$ -adapted Poisson random measure associated with a Lévy measure  $\nu$  defined on  $\mathbb{R}^n \setminus \{0\}$  and satisfying  $\int_{\mathbb{R}^n \setminus \{0\}} (|y|^2 \wedge 1) \nu(dy) < \infty$  (see, e.g., [3]).  $\tilde{N}(dt, dy) = N(dt, dy) - \nu(dy)dt$  is the compensated  $\{\mathfrak{F}_t\}_{t \geq 0}$ -martingale measure, where both  $N$  and  $\tilde{N}$  are independent of the Brownian motion  $W$ . In Eq (1.2),  $dE_t$  describes the drift of the time-changed stochastic process and  $\tilde{N}(dE_t, dz)$  is the compensated Poisson random measure that captures the jumps in addition to the time-changed Brownian motion. We define the product probability space by  $(\Omega, \mathfrak{F}, \mathbb{P}) := (\Omega_W \times \Omega_{\mathcal{D}}, \mathfrak{F}^W \times \mathfrak{F}^{\mathcal{D}}, \mathbb{P}_W \times \mathbb{P}_{\mathcal{D}})$ . Let  $\mathbb{E}$  denotes the expectation under the probability measure  $\mathbb{P}$ ,  $\mathbb{E}_{\mathcal{D}}$  denotes the expectation under the probability measure  $\mathbb{P}_{\mathcal{D}}$ , and  $\mathbb{E}_W$  refers to the expectation under the probability measure  $\mathbb{P}_W$ . It is shown that  $\mathbb{E}(\cdot) = \mathbb{E}_{\mathcal{D}}\mathbb{E}_W(\cdot) = \mathbb{E}_W\mathbb{E}_{\mathcal{D}}(\cdot)$ . Assume that  $f(t, E_t, 0, \delta_0) = g(t, E_t, 0, \delta_0) = \tilde{h}(t, E_t, 0, \delta_0, z) = 0$  for all  $0 \leq t, E_t < \infty$ , where  $\delta_0$  is the Dirac measure at 0.

The following lemma is borrowed from [16].

**Lemma 2.1.** *Suppose  $\alpha(t)$  and  $\lambda(t)$  are integrable and  $\mathcal{G}_t$ -measurable. Then, with probability one,*

$$\int_0^t \alpha(\tau) dE_\tau + \int_0^t \lambda(\tau) dW_\tau = \int_0^{E_t} \alpha(\mathcal{D}(\tau-)) d\tau + \int_0^{E_t} \lambda(\mathcal{D}(\tau-)) dW_\tau, \quad t \geq 0,$$

where the  $\mathcal{G}_t$ -measurable time-change  $E_t$  is the general inverse of  $\beta$ -stable subordinator  $\mathcal{D}(t)$ .

For any  $p \geq 2$ , consider the following subspace of  $\mathcal{M}(\mathbb{R}^n)$ :

$$\mathcal{M}_p(\mathbb{R}^n) := \left\{ \alpha \in \mathcal{M}(\mathbb{R}^n) : \alpha(|\cdot|^p) := \int_{\mathbb{R}^n} |y|^p \alpha(dy) < \infty \right\},$$

which is a polish space under the  $L^p$ -Wasserstein distance

$$\rho_p(\alpha_1, \alpha_2) := \inf_{\pi \in \mathcal{L}(\alpha_1, \alpha_2)} \left( \int_{\mathbb{R}^n \times \mathbb{R}^n} |y_1 - y_2|^p \pi(dy_1, dy_2) \right)^{\frac{1}{p}}, \quad \alpha_1, \alpha_2 \in \mathcal{M}_p(\mathbb{R}^n),$$

where  $\mathcal{L}(\alpha_1, \alpha_2)$  is the set of probability measures on  $\mathbb{R}^{n \times n}$  whose marginals are, respectively,  $\alpha_1, \alpha_2$ . In the case of  $\alpha_1 = \mathcal{L}(Y)$ ,  $\alpha_2 = \mathcal{L}(Z)$ , the corresponding distribution of random variables  $Y$  and  $Z$ , respectively, is

$$\rho_p(\alpha_1, \alpha_2)^p \leq \mathbb{E}|Y - Z|^p.$$

Throughout the article, the letter  $C > 0$  will refer to a constant whose value differs from one place to another. The dependence of a constant on the parameters will be assigned if it is necessary. Let

$C(J; \mathbb{R}^n)$  be the collection of all  $\mathbb{R}^n$ -valued continuous functions on  $J$  carrying the supremum norm and  $\mathbb{S}^p(\Omega; C(J; \mathbb{R}^n))$  be the totality of  $C(J; \mathbb{R}^n)$ -valued random variables  $Y$  with  $\mathbb{E}(\sup_{0 \leq t \leq T} |Y(t)|^p) < \infty$ . Hence,  $\mathbb{S}^p(\Omega; C(J; \mathbb{R}^n))$  is a Banach space under the norm

$$\|Y\|_{\mathbb{S}^p} := (\mathbb{E}(\sup_{0 \leq t \leq T} |Y(t)|^p))^{\frac{1}{p}}.$$

The rest of this section is devoted to recalling the concept of the  $L$ -derivative for functions on  $\mathcal{M}_2(\mathbb{R}^n)$ . For any  $\alpha \in \mathcal{M}_2(\mathbb{R}^n)$ , set

$$\mathcal{L}^2(\mathbb{R}^n, \alpha; \mathbb{R}^n) := \left\{ \varphi : \mathbb{R}^n \mapsto \mathbb{R}^n; \varphi \text{ is measurable with } \alpha(|\varphi|^2) := \int_{\mathbb{R}^n} |\varphi(y)|^2 \alpha(dy) < \infty \right\}.$$

Let  $|\cdot|$  and  $\|\cdot\|$  refer to the vector and matrix norms on  $\mathbb{R}^n$ , respectively. Let  $\mathcal{I}$  be the identity map on  $\mathbb{R}^n$ . For  $\alpha \in \mathcal{M}_2(\mathbb{R}^n)$  and  $\varphi \in \mathcal{L}^2(\mathbb{R}^n, \alpha; \mathbb{R}^n)$ ,  $\alpha(\varphi) := \int_{\mathbb{R}^n} \varphi(y) \alpha(dy)$  and  $\alpha \circ (\mathcal{I} + \varphi)^{-1} \in \mathcal{M}_2(\mathbb{R}^n)$ . Let  $\mathbb{C}^{1,1,2}(J \times \mathbb{R}^+ \times \mathbb{R}^n)$  refer to the set of all functions  $F(t_1, t_2, y) : J \times \mathbb{R}^+ \times \mathbb{R}^n \rightarrow \mathbb{R}$ , which are continuously once differentiable in  $t_1$  and  $t_2$  as well as continuously twice differentiable in  $y$ .

The following definitions are taken from [38, 40].

**Definition 2.1.** (1) The function  $F : \mathcal{M}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  is called  $L$ -differentiable at  $\alpha \in \mathcal{M}_2(\mathbb{R}^n)$ , if the functional

$$\mathcal{L}^2(\mathbb{R}^n, \alpha; \mathbb{R}^n) \ni \varphi \rightarrow F(\alpha \circ (\mathcal{I} + \varphi)^{-1})$$

is Fréchet differentiable at  $0 \in \mathcal{L}^2(\mathbb{R}^n, \alpha; \mathbb{R}^n)$ ; i.e., there exists a unique  $\zeta \in \mathcal{L}^2(\mathbb{R}^n, \alpha; \mathbb{R}^n)$  such that

$$\lim_{\alpha(|\varphi|^2) \rightarrow 0} \frac{F(\alpha \circ (\mathcal{I} + \varphi)^{-1}) - F(\alpha) - \alpha(\langle \zeta, \varphi \rangle)}{\sqrt{\alpha(|\varphi|^2)}} = 0,$$

where  $\langle \cdot, \cdot \rangle$  is the inner product on  $\mathbb{R}^n$  and  $\alpha(\varphi) := \int_{\mathbb{R}^n} \varphi(y) \alpha(dy)$ . Let  $\partial_\alpha F(\alpha) := \zeta$  be the  $L$ -derivative of  $F$  at  $\alpha$ .

(2) A function  $F : \mathcal{M}_2(\mathbb{R}^n) \rightarrow \mathbb{R}$  is said to be  $L$ -differentiable on  $\mathcal{M}_2(\mathbb{R}^n)$  if the  $L$ -derivative  $\partial_\alpha F(\alpha)$  exists for all  $\alpha \in \mathcal{M}_2(\mathbb{R}^n)$ .

**Definition 2.2.** A function  $F : \mathcal{M}_2(\mathbb{R}^n) \mapsto \mathbb{R}$  is said to be in the space  $\mathbb{C}^2(\mathcal{M}_2(\mathbb{R}^n))$ , if  $F$  is  $L$ -differentiable on  $\mathcal{M}_2(\mathbb{R}^n)$ , and its derivative  $\partial_\alpha F(x) : \mathcal{M}_2(\mathbb{R}^n) \times \mathbb{R}^n \mapsto \mathbb{R}^n$  is continuous at every  $(\alpha, x)$  for  $x \in \text{supp}(\alpha)$ .  $\partial_\alpha F(x)$  is differentiable in  $x$  and  $\partial_x \partial_\alpha F : \mathcal{M}_2(\mathbb{R}^n) \times \mathbb{R}^n \mapsto \mathbb{R}^{n \times n}$  is continuous at all  $(\alpha, x)$  satisfying  $x \in \text{supp}(\alpha)$ . In addition,  $\partial_\alpha^2 F : \mathcal{M}_2(\mathbb{R}^n) \times \mathbb{R}^n \times \mathbb{R}^n \mapsto \mathbb{R}^{n \times n}$  exists and is continuous.

**Definition 2.3.** Assume that  $F : J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n) \mapsto \mathbb{R}$ . If  $F(\cdot, \cdot, \cdot, \alpha)$  is in  $\mathbb{C}^{1,1,2}(J \times \mathbb{R}^+ \times \mathbb{R}^n)$  for any  $\alpha$ , and  $F(t, E, Y, \cdot) \in \mathbb{C}^2(\mathbb{R}^n)$  for every  $(t, E, Y)$ , and all the partial derivatives

$$\begin{aligned} \partial_{t_1} F(t, E, Y, \alpha), \quad \partial_{t_2} F(t, E, Y, \alpha), \quad \partial_Y F(t, E, Y, \alpha), \quad \partial_Y^2 F(t, E, Y, \alpha), \\ \partial_\alpha F(t, E, Y, \alpha)(x), \quad \partial_x \partial_\alpha F(t, E, Y, \alpha)(x), \quad \partial_\alpha^2 F(t, E, Y, \alpha)(x, \hat{x}) \end{aligned}$$

are continuous with respect to  $(t, E, Y, \alpha)$ ,  $(t, E, Y, \alpha, x)$ , or  $(t, E, Y, \alpha, x, \hat{x})$ , and  $F$  is said to be in the set  $\mathbb{C}^{1,1,2,(2)}(J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$ .

**Definition 2.4.** A function  $F : J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n) \mapsto \mathbb{R}$  is said to be in  $\wp_b(J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$ , if  $F \in \mathbb{C}^{1,1,2,(2)}(J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$  and all its derivatives are uniformly bounded on  $J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n)$ . In addition, if  $F \in \wp_b(J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$  and  $F \geq 0$ , we say that  $F \in \wp_{b,+}(J \times \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$ .

### 3. Existence and uniqueness

This section is concerned with establishing the existence and uniqueness of strong solutions for Eq (1.2). In order to derive the results of this section, assume that the coefficients  $f(t_1, t_2, y, \alpha)$ ,  $g(t_1, t_2, y, \alpha)$ , and  $\tilde{h}(t_1, t_2, y, \alpha, z)$  satisfy the following conditions.

**Condition 3.1.** There exists a positive constant  $L > 0$  such that for any  $t_1 \in J$ ,  $t_2 \in \mathbb{R}^+$  and  $y_1, y_2 \in \mathbb{R}^n$ ,  $\alpha_1, \alpha_2 \in \mathcal{M}_2(\mathbb{R}^n)$ ,

$$|f(t_1, t_2, y_1, \alpha_1) - f(t_1, t_2, y_2, \alpha_2)|^2 + \|g(t_1, t_2, y_1, \alpha_1) - g(t_1, t_2, y_2, \alpha_2)\|^2 + \int_{|z|<c} |\tilde{h}(t_1, t_2, y_1, \alpha_1, z) - \tilde{h}(t_1, t_2, y_2, \alpha_2, z)|^2 \nu(dz) \leq L\Gamma(|y_1 - y_2|^2 + \rho_2(\alpha_1, \alpha_2)^2),$$

where  $\Gamma(\cdot) : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  is continuous, non-decreasing, and concave function such that  $\Gamma(0) = 0$ ,  $\Gamma(y) > 0$  for every  $y > 0$  and  $\int_{0^+} \frac{1}{\Gamma(y)} dy = +\infty$ .

**Condition 3.2.** There exists a positive constant  $L_1 > 0$  such that for any  $t_1 \in J$ ,  $t_2 \in \mathbb{R}^+$ ,  $y \in \mathbb{R}^n$ , and  $\alpha \in \mathcal{M}_2(\mathbb{R}^n)$ ,

$$|f(t_1, t_2, y, \alpha)|^2 + \|g(t_1, t_2, y, \alpha)\|^2 + \int_{|z|<c} |\tilde{h}(t_1, t_2, y, \alpha, z)|^2 \nu(dz) \leq L_1(1 + |y|^2 + \rho_2(\alpha, \delta_0)^2).$$

**Condition 3.3 (Technical condition).** For any right-continuous  $\mathcal{G}_t$ -adapted stochastic process  $Y(t)$  with left limits, we have

$$f(t, E_t, Y(t), \mathcal{L}(Y(t))), g(t, E_t, Y(t), \mathcal{L}(Y(t))), \tilde{h}(t, E_t, Y(t), \mathcal{L}(Y(t)), z) \in \mathbb{L}(\mathcal{G}_t),$$

where  $\mathbb{L}(\mathcal{G}_t)$  denotes the totality of càglàd (i.e., sample paths that are left continuous with right limits) and  $\mathcal{G}_t$ -adapted processes.

**Remark 3.1.** To demonstrate our results generality, we give some concrete examples of the function  $\Gamma(\cdot)$ . Let  $\acute{L} > 0$  and let  $0 < \epsilon < 1$  be sufficiently small. Define

$$\begin{aligned} \Gamma_1(y) &= \acute{L}y, & y \geq 0, \\ \Gamma_2(y) &= \begin{cases} y \log(y^{-1}), & 0 \leq y \leq \epsilon, \\ \epsilon \log(\epsilon^{-1}) + \acute{\Gamma}_2(\epsilon-)(y - \epsilon), & y > \epsilon, \end{cases} \\ \Gamma_3(y) &= \begin{cases} y \log(y^{-1}) \log(\log(y^{-1})), & 0 \leq y \leq \epsilon, \\ \epsilon \log(\epsilon^{-1}) \log(\log(\epsilon^{-1})) + \acute{\Gamma}_3(\epsilon-)(y - \epsilon), & y > \epsilon, \end{cases} \end{aligned}$$

where  $\acute{\Gamma}_2$  and  $\acute{\Gamma}_3$  denote the derivatives of functions  $\Gamma_2$  and  $\Gamma_3$ , respectively. They are all concave and non-decreasing functions satisfying  $\int_{0^+} \frac{1}{\Gamma_i(y)} dy = +\infty$ ,  $i = 1, 2, 3$ .

Now, we introduce the following time-changed retarded integral inequality, which is essential in this work.

**Lemma 3.1.** Assume  $\mathcal{D}(t)$  is a  $\beta$ -stable subordinator corresponding to the inverse stable subordinator  $E_t$ . Let  $T > 0$  and  $\Gamma : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  be concave continuous and nondecreasing functions such that  $\Gamma(t) > 0$

for all  $t > 0$ . The functions  $a, d : \Omega \times \mathbb{R}^+ \rightarrow \mathbb{R}^+$  are assumed to be  $\mathcal{G}_t$ -measurable and integrable with respect to  $E_t$ . Let  $r(t) > 0$  be a monotone and nondecreasing function. If

$$a(t) \leq r(t) + \int_0^t d(s)\Gamma(a(s))dE_s, \quad t \geq 0, \quad (3.1)$$

then

$$a(t) \leq G^{-1}\left(G(r(t)) + \int_0^t d(s)dE_s\right)$$

holds for all  $t \geq 0$  such that

$$G(r(t)) + \int_0^t d(s)dE_s \in \text{Dom}(G^{-1}),$$

where

$$G(t) = \int_0^t \frac{ds}{\Gamma(s)}, \quad t > 0,$$

and  $G^{-1}$  is the inverse function of  $G$ . In particular, if, moreover,  $a(0) = 0$  and  $\int_{0+} \frac{ds}{\Gamma(s)} = \infty$ , then  $a(t) = 0$ , for all  $t \in J$ .

*Proof.* Let

$$e(t) := r(t) + \int_0^t d(s)\Gamma(a(s))dE_s, \quad t \geq 0. \quad (3.2)$$

Since  $a(s)$  and  $d(s)$  are positive and  $r(t)$  is non-decreasing, the function  $e(t)$  is non-decreasing. Moreover, from Eqs (3.1) and (3.2),

$$a(t) \leq e(t) \quad \text{or} \quad \Gamma(a(t)) \leq \Gamma(e(t)), \quad t \geq 0,$$

which implies that

$$e(t) \leq r(t) + \int_0^t d(s)\Gamma(e(s))dE_s, \quad t \geq 0.$$

Applying Lemma 2.1, we have

$$e(t) \leq r(t) + \int_0^{E_t} d(\mathcal{D}(s-))\Gamma(e(\mathcal{D}(s-)))ds, \quad t \geq 0. \quad (3.3)$$

Actually, for  $0 \leq t \leq E_t$ ,  $\mathcal{D}(t-)$  is defined as

$$\mathcal{D}(t-) = \inf\{s : s \in [0, T], E_s \geq t\} \wedge T,$$

which means

$$E_{\mathcal{D}(t-)} = t \quad \text{and} \quad \mathcal{D}(E_t-) \leq t. \quad (3.4)$$

Let  $\tau \in [0, \infty)$ , and it holds from (3.3) and (3.4) that

$$\begin{aligned} e(\mathcal{D}(\tau-)) &\leq r(\mathcal{D}(\tau-)) + \int_0^{E_{\mathcal{D}(\tau-)}} d(\mathcal{D}(s-))\Gamma(e(\mathcal{D}(s-)))ds \\ &= r(\mathcal{D}(\tau-)) + \int_0^\tau d(\mathcal{D}(s-))\Gamma(e(\mathcal{D}(s-)))ds. \end{aligned}$$

We apply a generalization of Bihari's inequality [5] and Lemma 3.6 in [25] path by path to yield

$$a(\mathcal{D}(\tau-)) \leq e(\mathcal{D}(\tau-)) \leq G^{-1}\left(G(r(\mathcal{D}(\tau-))) + \int_0^\tau d(\mathcal{D}(s-))ds\right).$$

For any  $t \geq 0$ , we can see that there is  $\tau \in \mathbb{R}^+$  such that  $\mathcal{D}(\tau-) = t$  provided that  $E$  is strictly increasing in some neighborhood of  $t$ . As  $r(t)$  is non-decreasing, employing (3.4) and Lemma 2.1, we gain

$$\begin{aligned} a(t) &\leq e(t) \leq G^{-1}\left(G(r(t)) + \int_0^{E_t} d(\mathcal{D}(s-))ds\right) \\ &:= G^{-1}\left(G(r(t)) + \int_0^t d(s)dE_s\right). \end{aligned}$$

If  $E$  is a constant  $\tau$  in some neighborhood  $U(t; \epsilon)$  of  $t$ , then  $t - \epsilon = \mathcal{D}(\tau-)$ . Employing (3.4) and since  $r(t)$  is non-decreasing, we have

$$a(t - \epsilon) \leq e(t - \epsilon) \leq G^{-1}\left(G(r(t - \epsilon)) + \int_0^{E_{t-\epsilon}} d(\mathcal{D}(s-))ds\right).$$

Then, for any  $t \geq 0$ , we gain by Lemma 2.1

$$a(t) \leq e(t) \leq G^{-1}\left(G(r(t)) + \int_0^t d(s)dE_s\right).$$

Hence, the proof is complete.  $\square$

Now, we construct the Carathéodory approximation as follows. For any integer  $k \geq 1$ , define  $Y^k(t) = Y_0$  for all  $-1 \leq t \leq 0$  and

$$\begin{aligned} Y^k(t) &= Y_0 + \int_0^t f(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})))dE_s \\ &\quad + \int_0^t g(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})))dW_{E_s} \\ &\quad + \int_0^t \int_{|z|<c} \tilde{h}(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})), z)\tilde{N}(dE_s, dz), \quad \forall t \geq 0. \end{aligned} \quad (3.5)$$

Note that for  $0 \leq t \leq \frac{1}{k}$ ,  $Y^k(t)$  can be computed by

$$Y^k(t) = Y_0 + \int_0^t f(s, E_s, Y_0, \mathcal{L}(Y_0))dE_s$$

$$\begin{aligned}
& + \int_0^t g(s, E_s, Y_0, \mathcal{L}(Y_0)) dW_{E_s} \\
& + \int_0^t \int_{|z|<c} \tilde{h}(s, E_s, Y_0, \mathcal{L}(Y_0), z) \tilde{N}(dE_s, dz).
\end{aligned}$$

Then for  $\frac{1}{k} < t < \frac{2}{k}$ ,

$$\begin{aligned}
Y^k(t) & = Y^k\left(\frac{1}{k}\right) + \int_{\frac{1}{k}}^t f(s, E_s, Y^k\left(s - \frac{1}{k}\right), \mathcal{L}(Y^k\left(s - \frac{1}{k}\right))) dE_s \\
& + \int_{\frac{1}{k}}^t g(s, E_s, Y^k\left(s - \frac{1}{k}\right), \mathcal{L}(Y^k\left(s - \frac{1}{k}\right))) dW_{E_s} \\
& + \int_{\frac{1}{k}}^t \int_{|z|<c} \tilde{h}(s, E_s, Y^k\left(s - \frac{1}{k}\right), \mathcal{L}(Y^k\left(s - \frac{1}{k}\right)), z) \tilde{N}(dE_s, dz)
\end{aligned}$$

and so on. In other words,  $Y^k(t)$  can be computed step by step on the intervals  $[0, \frac{1}{k}]$ ,  $[\frac{1}{k}, \frac{2}{k}]$ , ....

Next, we will show that the sequence of stochastic processes  $\{Y^k(t)\}_{k \geq 1}$  given in (3.5) is uniformly bounded.

**Lemma 3.2.** *Assume Conditions 3.2 and 3.3 hold. Then, for all  $k \geq 1$  and  $t \in J$ ,  $\{Y^k(t)\}_{k \geq 1}$  is uniformly bounded with the property*

$$\mathbb{E}(\sup_{0 \leq s \leq T} |Y^k(s)|^2) \leq \infty.$$

*Proof.* It follows from Eq (3.5) that

$$\begin{aligned}
& \mathbb{E}_W(\sup_{0 \leq s \leq t} |Y^k(s)|^2) \\
& \leq 4\mathbb{E}_W|Y_0|^2 + 4\mathbb{E}_W\left(\sup_{0 \leq s \leq t} \left| \int_0^s f(v, E_v, Y^k\left(v - \frac{1}{k}\right), \mathcal{L}(Y^k\left(v - \frac{1}{k}\right))) dE_v \right|^2\right) \\
& + 4\mathbb{E}_W\left(\sup_{0 \leq s \leq t} \left| \int_0^s g(v, E_v, Y^k\left(v - \frac{1}{k}\right), \mathcal{L}(Y^k\left(v - \frac{1}{k}\right))) dW_{E_v} \right|^2\right) \\
& + 4\mathbb{E}_W\left(\sup_{0 \leq s \leq t} \left| \int_0^s \int_{|z|<c} \tilde{h}(v, E_v, Y^k\left(v - \frac{1}{k}\right), \mathcal{L}(Y^k\left(v - \frac{1}{k}\right)), z) \tilde{N}(dE_v, dz) \right|^2\right) \\
& := 4\mathbb{E}_W|Y_0|^2 + I_1 + I_2 + I_3.
\end{aligned} \tag{3.6}$$

For  $I_1$ , by Hölder's inequality and Condition 3.2, we gain

$$\begin{aligned}
I_1 & = 4\mathbb{E}_W\left(\sup_{0 \leq s \leq t} \left| \int_0^s f(v, E_v, Y^k\left(v - \frac{1}{k}\right), \mathcal{L}(Y^k\left(v - \frac{1}{k}\right))) dE_v \right|^2\right) \\
& \leq 4E_T \mathbb{E}_W \int_0^t |f(s, E_s, Y^k\left(s - \frac{1}{k}\right), \mathcal{L}(Y^k\left(s - \frac{1}{k}\right)))|^2 dE_s \\
& \leq 4E_T \mathbb{E}_W \int_0^t L_1 \left(1 + |Y^k\left(s - \frac{1}{k}\right)|^2 + \rho_2(\mathcal{L}(Y^k\left(s - \frac{1}{k}\right)), \delta_0)^2\right) dE_s
\end{aligned}$$

$$\begin{aligned}
&\leq 4E_T L_1 \int_0^t \left(1 + 2\mathbb{E}_W |Y^k(s - \frac{1}{k})|^2\right) dE_s \\
&\leq 4E_T L_1 \int_0^t \left(1 + 2\mathbb{E}_W(\sup_{0 \leq v \leq s} |Y^k(v-)|^2)\right) dE_s.
\end{aligned} \tag{3.7}$$

For  $I_2$ , by the Burkholder-Davis-Gundy inequality [12] and Condition 3.2, we obtain

$$\begin{aligned}
I_2 &= 4\mathbb{E}_W \left( \sup_{0 \leq s \leq t} \left| \int_0^s g(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k}))) dW_{E_v} \right|^2 \right) \\
&\leq 4b_2 \mathbb{E}_W \int_0^t \|g(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})))\|^2 dE_s \\
&\leq 4b_2 \mathbb{E}_W \int_0^t L_1 \left(1 + |Y^k(s - \frac{1}{k})|^2 + \rho_2(\mathcal{L}(Y^k(s - \frac{1}{k})), \delta_0)^2\right) dE_s \\
&\leq 4b_2 L_1 \int_0^t \left(1 + 2\mathbb{E}_W |Y^k(s - \frac{1}{k})|^2\right) dE_s \\
&\leq 4b_2 L_1 \int_0^t \left(1 + 2\mathbb{E}_W(\sup_{0 \leq v \leq s} |Y^k(v-)|^2)\right) dE_s,
\end{aligned} \tag{3.8}$$

where  $b_2 > 0$  comes from [12]. For  $I_3$ , by Doob's martingale inequality, Itô's isometry (comes from [39]), and Condition 3.2, we gain

$$\begin{aligned}
I_3 &= 4\mathbb{E}_W \left( \sup_{0 \leq s \leq t} \left| \int_0^s \int_{|z| < c} \tilde{h}(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})), z) \tilde{N}(dE_v, dz) \right|^2 \right) \\
&\leq 16\mathbb{E}_W \left| \int_0^t \int_{|z| < c} \tilde{h}(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})), z) \tilde{N}(dE_s, dz) \right|^2 \\
&\leq 16\mathbb{E}_W \int_0^t \int_{|z| < c} |\tilde{h}(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})), z)|^2 \nu(dz) dE_s \\
&\leq 16\mathbb{E}_W \int_0^t L_1 \left(1 + |Y^k(s - \frac{1}{k})|^2 + \rho_2(\mathcal{L}(Y^k(s - \frac{1}{k})), \delta_0)^2\right) dE_s \\
&\leq 16L_1 \int_0^t \left(1 + 2\mathbb{E}_W |Y^k(s - \frac{1}{k})|^2\right) dE_s \\
&\leq 16L_1 \int_0^t \left(1 + 2\mathbb{E}_W(\sup_{0 \leq v \leq s} |Y^k(v-)|^2)\right) dE_s.
\end{aligned} \tag{3.9}$$

Then, from Eqs (3.6)–(3.9), we gain

$$\begin{aligned}
1 + 2\mathbb{E}_W(\sup_{0 \leq s \leq t} |Y^k(s)|^2) &\leq 1 + 8\mathbb{E}_W |Y_0|^2 + 8L_1(E_T + b_2 + 4) \\
&\quad \times \int_0^t \left(1 + 2\mathbb{E}_W(\sup_{0 \leq v \leq s} |Y^k(v-)|^2)\right) dE_s \\
&\leq 1 + 8\mathbb{E}_W |Y_0|^2 + C_{E_T} \int_0^t \left(1 + 2\mathbb{E}_W(\sup_{0 \leq v \leq s} |Y^k(v-)|^2)\right) dE_s,
\end{aligned}$$

which, with the help of the time-changed Gronwall inequality [12], gives

$$\mathbb{E}_W(\sup_{0 \leq s \leq t} |Y^k(s)|^2) \leq (1 + 8\mathbb{E}_W |Y_0|^2) e^{C_{E_T} E_T}. \tag{3.10}$$

Taking  $\mathbb{E}_{\mathcal{D}}$  for both sides of (3.10) and since  $\alpha(0, \infty) = \infty$  guarantees that the inverse  $E$  of  $\mathcal{D}$  has a finite exponential moment, we have

$$\mathbb{E}(\sup_{0 \leq s \leq T} |Y^k(s)|^2) \leq (1 + 8\mathbb{E}|Y_0|^2)\mathbb{E}(e^{C_{E_T} E_T}) < \infty,$$

where  $C_{E_T} = 8L_1(E_T + b_2 + 4)$ .  $\square$

**Lemma 3.3.** *Assume Conditions 3.2 and 3.3 hold. Then*

$$\mathbb{E}_{\mathbb{W}}|Y^k(t) - Y^k(s)|^2 \leq C_{b_2, E_T}(E_t - E_s), \quad t \geq s,$$

where  $C_{b_2, E_T}$  is a positive constant depending on  $b_2$  and  $E_T$ .

*Proof.* By the Eq (3.5), we have

$$\begin{aligned} & \mathbb{E}_{\mathbb{W}}|Y^k(t) - Y^k(s)|^2 \\ & \leq 3\mathbb{E}_{\mathbb{W}} \left| \int_s^t f(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k}))) dE_v \right|^2 \\ & \quad + 3\mathbb{E}_{\mathbb{W}} \left| \int_s^t g(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k}))) dW_{E_v} \right|^2 \\ & \quad + 3\mathbb{E}_{\mathbb{W}} \left| \int_s^t \int_{|z| < c} \tilde{h}(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})), z) \tilde{N}(dE_v, dz) \right|^2 \\ & := \gamma_1 + \gamma_2 + \gamma_3. \end{aligned} \tag{3.11}$$

By Hölder's inequality and Condition 3.2, we get

$$\begin{aligned} \gamma_1 & \leq 3(E_t - E_s)\mathbb{E}_{\mathbb{W}} \int_s^t |f(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})))|^2 dE_v \\ & \leq 3(E_t - E_s)\mathbb{E}_{\mathbb{W}} \int_s^t L_1 \left( 1 + |Y^k(v - \frac{1}{k})|^2 + \rho_2(\mathcal{L}(Y^k(v - \frac{1}{k})), \delta_0)^2 \right) dE_v \\ & \leq 3L_1(E_T - E_s) \int_s^t \left( 1 + 2\mathbb{E}_{\mathbb{W}}|Y^k(v - \frac{1}{k})|^2 \right) dE_v \\ & \leq 3L_1(E_T - E_s) \int_s^t \left( 1 + 2\mathbb{E}_{\mathbb{W}}(\sup_{0 \leq v_1 \leq v} |Y^k(v_1 - \frac{1}{k})|^2) \right) dE_v. \end{aligned} \tag{3.12}$$

For  $\gamma_2$ , by the Burkholder-Davis-Gundy inequality [12] and Condition 3.2, we have

$$\begin{aligned} \gamma_2 & \leq 3b_2\mathbb{E}_{\mathbb{W}} \int_s^t \|g(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})))\|^2 dE_v \\ & \leq 3b_2\mathbb{E}_{\mathbb{W}} \int_s^t L_1 \left( 1 + |Y^k(v - \frac{1}{k})|^2 + \rho_2(\mathcal{L}(Y^k(v - \frac{1}{k})), \delta_0)^2 \right) dE_v \\ & \leq 3b_2L_1 \int_s^t \left( 1 + 2\mathbb{E}_{\mathbb{W}}|Y^k(v - \frac{1}{k})|^2 \right) dE_v \\ & \leq 3b_2L_1 \int_s^t \left( 1 + 2\mathbb{E}_{\mathbb{W}}(\sup_{0 \leq v_1 \leq v} |Y^k(v_1 - \frac{1}{k})|^2) \right) dE_v. \end{aligned} \tag{3.13}$$

For  $\gamma_3$ , by Doob's martingale inequality, Itô isometry (from [39]), and Condition 3.2, we gain

$$\begin{aligned}
 \gamma_3 &\leq 12\mathbb{E}_W \int_s^t \int_{|z|<c} |\tilde{h}(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})), z)|^2 \nu(dz) dE_v \\
 &\leq 12\mathbb{E}_W \int_s^t L_1 \left( 1 + |Y^k(v - \frac{1}{k})|^2 + \rho_2(\mathcal{L}(Y^k(v - \frac{1}{k})), \delta_0)^2 \right) dE_v \\
 &\leq 12L_1 \int_s^t \left( 1 + 2\mathbb{E}_W |Y^k(v - \frac{1}{k})|^2 \right) dE_v \\
 &\leq 12L_1 \int_s^t \left( 1 + 2\mathbb{E}_W (\sup_{0 \leq v_1 \leq v} |Y^k(v_1 -)|^2) \right) dE_v.
 \end{aligned} \tag{3.14}$$

Thus, by Eqs (3.11)–(3.14) and Lemma 3.2, we gain

$$\begin{aligned}
 \mathbb{E}_W |Y^k(t) - Y^k(s)|^2 &\leq 3L_1(E_T - E_s + b_2 + 4) \\
 &\quad \times \int_s^t \left( 1 + 2\mathbb{E}_W (\sup_{0 \leq v_1 \leq v} |Y^k(v_1 -)|^2) \right) dE_v \\
 &\leq C_{b_2, E_T}(E_t - E_s),
 \end{aligned}$$

where  $C_{b_2, E_T} = 3L_1(E_T - E_s + b_2 + 4)(1 + 2\mathbb{E}_W(\sup_{0 \leq v_1 \leq v} |Y^k(v_1 -)|^2))$ .  $\square$

**Theorem 3.1.** Assume Conditions 3.1–3.3 hold. Then, for any initial value  $Y_0$  satisfying  $\mathbb{E}|Y_0|^2 < \infty$ , the Eq (1.2) has a unique solution  $Y(t)$ ,  $t \in J$ .

*Proof.* We will split the proof into the following two parts.

**First, we show the existence.**

In fact, for  $k > l \geq 1$ , it is routine to obtain

$$\begin{aligned}
 &\mathbb{E}_W (\sup_{0 \leq s \leq t} |Y^k(s) - Y^l(s)|^2) \\
 &\leq 3\mathbb{E}_W \left( \sup_{0 \leq s \leq t} \left| \int_0^s (f(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k}))) \right. \right. \\
 &\quad \left. \left. - f(v, E_v, Y^l(v - \frac{1}{l}), \mathcal{L}(Y^l(v - \frac{1}{l}))) \right) dE_v \right|^2 \\
 &\quad + 3\mathbb{E}_W \left( \sup_{0 \leq s \leq t} \left| \int_0^s (g(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k}))) \right. \right. \\
 &\quad \left. \left. - g(v, E_v, Y^l(v - \frac{1}{l}), \mathcal{L}(Y^l(v - \frac{1}{l}))) \right) dW_{E_v} \right|^2 \\
 &\quad + 3\mathbb{E}_W \left( \sup_{0 \leq s \leq t} \left| \int_0^s \int_{|z|<c} (\tilde{h}(v, E_v, Y^k(v - \frac{1}{k}), \mathcal{L}(Y^k(v - \frac{1}{k})), z) \right. \right. \\
 &\quad \left. \left. - \tilde{h}(v, E_v, Y^l(v - \frac{1}{l}), \mathcal{L}(Y^l(v - \frac{1}{l})), z)) \tilde{N}(dE_v, dz) \right|^2 \right) \\
 &:= J_1 + J_2 + J_3.
 \end{aligned}$$

By Hölder's inequality, Condition 3.1, Jensen's inequality and Lemma 3.3, we gain

$$J_1 \leq 3E_T \mathbb{E}_W \int_0^t |f(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k})))|^2 ds$$

$$\begin{aligned}
& -f(s, E_s, Y^l(s - \frac{1}{l}), \mathcal{L}(Y^l(s - \frac{1}{l})))|^2 dE_s \\
\leq & 6E_T \mathbb{E}_W \int_0^t |f(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k}))) \\
& -f(s, E_s, Y^l(s - \frac{1}{l}), \mathcal{L}(Y^l(s - \frac{1}{l})))|^2 dE_s \\
& + 6E_T \mathbb{E}_W \int_0^t |f(s, E_s, Y^l(s - \frac{1}{k}), \mathcal{L}(Y^l(s - \frac{1}{k}))) \\
& -f(s, E_s, Y^l(s - \frac{1}{l}), \mathcal{L}(Y^l(s - \frac{1}{l})))|^2 dE_s \\
\leq & 6E_T L \mathbb{E}_W \int_0^t \Gamma \left( |Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2 \right. \\
& \left. + \rho_2(\mathcal{L}(Y^k(s - \frac{1}{k})), \mathcal{L}(Y^l(s - \frac{1}{k})))^2 \right) dE_s \\
& + 6E_T L \mathbb{E}_W \int_0^t \Gamma \left( |Y^l(s - \frac{1}{k}) - Y^l(s - \frac{1}{l})|^2 \right. \\
& \left. + \rho_2(\mathcal{L}(Y^l(s - \frac{1}{k})), \mathcal{L}(Y^l(s - \frac{1}{l})))^2 \right) dE_s \\
\leq & 6E_T L \int_0^t \Gamma \left( 2\mathbb{E}_W |Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2 \right) dE_s \\
& + 6E_T L \int_0^t \Gamma \left( 2\mathbb{E}_W |Y^l(s - \frac{1}{k}) - Y^l(s - \frac{1}{l})|^2 \right) dE_s \\
\leq & 6E_T L \int_0^t \Gamma \left( 2\mathbb{E}_W |Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2 \right) dE_s \\
& + 6E_T L \int_0^t \Gamma \left( 2C_{b_2, E_T} (E_{s-\frac{1}{k}} - E_{s-\frac{1}{l}}) \right) dE_s.
\end{aligned}$$

For  $J_2$ , by Burkholder-Davis-Gundy inequality [12], Condition 3.1, Jensen inequality and Lemma 3.3, we obtain

$$\begin{aligned}
J_2 & \leq 3b_2 \mathbb{E}_W \int_0^t \|g(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k}))) \\
& -g(s, E_s, Y^l(s - \frac{1}{l}), \mathcal{L}(Y^l(s - \frac{1}{l})))\|^2 dE_s \\
& \leq 6b_2 \mathbb{E}_W \int_0^t \|g(s, E_s, Y^k(s - \frac{1}{k}), \mathcal{L}(Y^k(s - \frac{1}{k}))) \\
& -g(s, E_s, Y^l(s - \frac{1}{k}), \mathcal{L}(Y^l(s - \frac{1}{k})))\|^2 dE_s \\
& + 6b_2 \mathbb{E}_W \int_0^t \|g(s, E_s, Y^l(s - \frac{1}{k}), \mathcal{L}(Y^l(s - \frac{1}{k}))) \\
& -g(s, E_s, Y^l(s - \frac{1}{l}), \mathcal{L}(Y^l(s - \frac{1}{l})))\|^2 dE_s \\
& \leq 6b_2 L \mathbb{E}_W \int_0^t \Gamma \left( |Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2 \right)
\end{aligned}$$

$$\begin{aligned}
& +\rho_2(\mathcal{L}(Y^k(s-\frac{1}{k})), \mathcal{L}(Y^l(s-\frac{1}{l})))^2)dE_s \\
& +6b_2L\mathbb{E}_W \int_0^t \Gamma\left(|Y^l(s-\frac{1}{k}) - Y^l(s-\frac{1}{l})|^2\right. \\
& \left. +\rho_2(\mathcal{L}(Y^l(s-\frac{1}{k})), \mathcal{L}(Y^l(s-\frac{1}{l})))^2\right)dE_s \\
\leq & 6b_2L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^k(s-\frac{1}{k}) - Y^l(s-\frac{1}{k})|^2\right)dE_s \\
& +6b_2L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^l(s-\frac{1}{k}) - Y^l(s-\frac{1}{l})|^2\right)dE_s \\
\leq & 6b_2L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^k(s-\frac{1}{k}) - Y^l(s-\frac{1}{k})|^2\right)dE_s \\
& +6b_2L \int_0^t \Gamma\left(2C_{b_2, E_T}(E_{s-\frac{1}{k}} - E_{s-\frac{1}{l}})\right)dE_s.
\end{aligned}$$

For  $J_3$ , by Doob's martingale inequality, Itô's isometry (from [39]), Condition 3.1, Jensen's inequality and Lemma 3.3, we have

$$\begin{aligned}
J_3 & \leq 12\mathbb{E}_W \int_0^t \int_{|z|<c} |\tilde{h}(s, E_s, Y^k(s-\frac{1}{k}), \mathcal{L}(Y^k(s-\frac{1}{k})), z) \\
& -\tilde{h}(s, E_s, Y^l(s-\frac{1}{l}), \mathcal{L}(Y^l(s-\frac{1}{l})), z)|^2 \nu(dz)dE_s \\
& \leq 24\mathbb{E}_W \int_0^t \int_{|z|<c} |\tilde{h}(s, E_s, Y^k(s-\frac{1}{k}), \mathcal{L}(Y^k(s-\frac{1}{k})), z) \\
& -\tilde{h}(s, E_s, Y^l(s-\frac{1}{k}), \mathcal{L}(Y^l(s-\frac{1}{k})), z)|^2 \nu(dz)dE_s \\
& +24\mathbb{E}_W \int_0^t \int_{|z|<c} |\tilde{h}(s, E_s, Y^l(s-\frac{1}{k}), \mathcal{L}(Y^l(s-\frac{1}{k})), z) \\
& -\tilde{h}(s, E_s, Y^l(s-\frac{1}{l}), \mathcal{L}(Y^l(s-\frac{1}{l})), z)|^2 \nu(dz)dE_s \\
& \leq 24L\mathbb{E}_W \int_0^t \Gamma\left(|Y^k(s-\frac{1}{k}) - Y^l(s-\frac{1}{k})|^2\right. \\
& \left. +\rho_2(\mathcal{L}(Y^k(s-\frac{1}{k})), \mathcal{L}(Y^l(s-\frac{1}{k})))^2\right)dE_s \\
& +24L\mathbb{E}_W \int_0^t \Gamma\left(|Y^l(s-\frac{1}{k}) - Y^l(s-\frac{1}{l})|^2\right. \\
& \left. +\rho_2(\mathcal{L}(Y^l(s-\frac{1}{k})), \mathcal{L}(Y^l(s-\frac{1}{l})))^2\right)dE_s \\
& \leq 24L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^k(s-\frac{1}{k}) - Y^l(s-\frac{1}{k})|^2\right)dE_s \\
& +24L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^l(s-\frac{1}{k}) - Y^l(s-\frac{1}{l})|^2\right)dE_s.
\end{aligned}$$

$$\begin{aligned} &\leq 24L \int_0^t \Gamma\left(2\mathbb{E}_W|Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2\right)dE_s \\ &\quad + 24L \int_0^t \Gamma\left(2C_{b_2, E_T}(E_{s-\frac{1}{k}} - E_{s-\frac{1}{l}})\right)dE_s. \end{aligned}$$

Hence, we gain

$$\begin{aligned} &\mathbb{E}_W(\sup_{0 \leq s \leq t} |Y^k(s) - Y^l(s)|^2) \\ &\leq 6L(E_T + b_2 + 4) \int_0^t \Gamma\left(2\mathbb{E}_W|Y^k(s - \frac{1}{k}) - Y^l(s - \frac{1}{k})|^2\right)dE_s \\ &\quad + 6L(E_T + b_2 + 4)E_T \Gamma\left(2C_{b_2, E_T}(E_{s-\frac{1}{k}} - E_{s-\frac{1}{l}})\right). \end{aligned}$$

By the definition of  $E_t$ , and from [48] and Theorem 2 of [22], we have

$$|E_{s-\frac{1}{k}} - E_{s-\frac{1}{l}}| \leq k_1 \left| \frac{1}{l} - \frac{1}{k} \right|. \quad (3.15)$$

Noticing (3.15), taking the limit as  $k, l \rightarrow \infty$  and using the fact that  $\Gamma(0) = 0$ , we gain, for every  $\varepsilon > 0$ ,

$$\begin{aligned} Q(t) &\leq 6L(E_T + b_2 + 4) \int_0^t \Gamma(2Q(s))dE_s \\ &\leq \varepsilon + 6L(E_T + b_2 + 4) \int_0^t \Gamma(2Q(s))dE_s, \end{aligned}$$

where  $Q(t) = \lim_{k, l \rightarrow \infty} \mathbb{E}_W(\sup_{0 \leq s \leq t} |Y^k(s) - Y^l(s)|^2)$ . Hence, Lemma 3.1 yields

$$Q(t) \leq \frac{1}{2}G^{-1}\left(G(2\varepsilon) + 12L(E_T + b_2 + 4)E_T\right),$$

where  $G(2\varepsilon) + 12L(E_T + b_2 + 4)E_T \in \text{Dom}(G^{-1})$ ,  $G^{-1}$  is the inverse function of  $G(\cdot)$ , and

$$G(\tau) = \int_1^\tau \frac{ds}{\Gamma(s)}, \quad \tau > 0.$$

By Condition 3.1, one sees that  $\lim_{\varepsilon \downarrow 0} G(\varepsilon) = -\infty$  and  $\text{Dom}(G^{-1}) = (-\infty, G(\infty))$ . Taking  $\mathbb{E}_D$  and letting  $\varepsilon \rightarrow 0$  gives  $Q(t) = 0$ , i.e.,

$$\mathbb{E}(\sup_{0 \leq s \leq t} |Y^k(s) - Y^l(s)|^2) \rightarrow 0, \quad \text{as } k, l \rightarrow \infty. \quad (3.16)$$

Consequently,  $\{Y^k(t)\}_{k \geq 1}$  is a Cauchy sequence in  $\mathbb{S}^2(\Omega; C(J; \mathbb{R}^n))$ , and then the limit is denoted by  $Y(t)$ . Therefore, allowing  $l \rightarrow \infty$  in (3.16), we conclude

$$\lim_{k \rightarrow \infty} \mathbb{E}(\sup_{0 \leq s \leq T} |Y^k(s) - Y(s)|^2) = 0. \quad (3.17)$$

Now, we will show that  $Y(t)$  is a solution to Eq (1.2). For all  $t \in J$ , we gain

$$\mathbb{E}|Y(t) - Y^k(t - \frac{1}{k})|^2 \leq 2\mathbb{E}|Y(t) - Y^k(t)|^2 + 2\mathbb{E}|Y^k(t) - Y^k(t - \frac{1}{k})|^2,$$

and using Lemma 3.3 and (3.17), we obtain  $\lim_{k \rightarrow \infty} \mathbb{E}|Y(t) - Y^k(t - \frac{1}{k})|^2 = 0$ . Hence, as  $k \rightarrow \infty$  in (3.5), we obtain

$$\begin{aligned} Y(t) &= Y_0 + \int_0^t f(s, E_s, Y(s), \mathcal{L}(Y(s)))dE_s \\ &\quad + \int_0^t g(s, E_s, Y(s), \mathcal{L}(Y(s)))dW_{E_s} \\ &\quad + \int_0^t \int_{|z|<c} \tilde{h}(s, E_s, Y(s), \mathcal{L}(Y(s)), z)\tilde{N}(dE_s, dz), \quad 0 \leq t \leq T, \end{aligned}$$

which proves that  $Y(t)$  is a solution for Eq (1.2). Therefore, the existence proof is finished.

**Next, we show the uniqueness.**

Let  $X(t)$  and  $Y(t)$  be two solutions for Eq (1.2) with  $X(0) = Y(0)$ . Then, using the same steps as above, we have

$$\begin{aligned} \mathbb{E}_W(\sup_{0 \leq s \leq t} |X(s) - Y(s)|^2) &\leq 3L(E_T + b_2 + 4) \\ &\quad \times \int_0^t \Gamma\left(2\mathbb{E}_W(\sup_{0 \leq v \leq s} |X(v-) - Y(v-)|^2)\right)dE_s, \end{aligned}$$

for all  $t \in J$ . Then Lemma 3.1 gives  $\mathbb{E}(\sup_{0 \leq s \leq t} |X(s) - Y(s)|^2) = 0$ , which implies that  $X(t) = Y(t)$ ,  $0 \leq t \leq T$   $\mathbb{P}$ -a.s. Thus, the proof of Theorem 3.1 is complete.  $\square$

**Remark 3.1.** Putting  $\tilde{h} \equiv 0$  in Eq (1.2), it reduces to the MFSDEs driven by time-changed Brownian motion studied by [21, 40]. Hence, Theorem 3.1 is still new for MFSDEs driven by time-changed Brownian motion. Then, our results generalize and improve those obtained in [21, 40].

#### 4. The generalized Itô formula

Nane and Ni [31] established an Itô formula for SDEs driven by time-changed Lévy noise. In this section, the Itô's formula will be extended from SDEs to MFSDEs driven by time-changed Lévy noise under Conditions 3.1–3.3.

**Lemma 4.1.** Assume Conditions 3.1–3.3 hold, and the function  $F : \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n) \mapsto \mathbb{R}$  belongs to  $\wp_b(\mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$ . Then, for any  $\mathfrak{F}_0$ -measurable random variable  $y_0$  satisfying  $\mathbb{E}|y_0|^2 < \infty$ , the following Itô formula holds:

$$\begin{aligned} &F(E_t, Y(t), \alpha_t) - F(0, y_0, \alpha_0) \\ &= \int_0^t \mathbb{L}_{f,g,\tilde{h}}F(E_s, Y(s-), \alpha) dE_s \\ &\quad + \int_0^t F_y(E_s, Y(s-), \alpha_0)g(s, E_s, Y(s-), \mathcal{L}(Y(s-)))dW_{E_s} \\ &\quad + \int_0^t \int_{|z|<c} [F(E_s, Y(s-) + \tilde{h}(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z), \alpha_0) \\ &\quad \quad - F(E_s, Y(s-), \alpha_0)]\tilde{N}(dz, dE_s), \end{aligned} \tag{4.1}$$

with

$$\mathbb{L}_{f,g,\tilde{h}}F(t_2, y, \alpha) = \partial_{t_2}F(t_2, y, \alpha_0) + F_y(t_2, y, \alpha_0)f(t, t_2, y, \alpha)$$

$$\begin{aligned}
& + \frac{1}{2} \operatorname{tr}[F_{yy}(t_2, y, \alpha_0)(gg^T)(t, t_2, y, \alpha)] \\
& + \partial_\alpha F(E_{t_1}, Y(E_{t_1}), \alpha)(y)f(t, t_2, y, \alpha) \\
& + \frac{1}{2} \operatorname{tr}[\partial_\nu \partial_\alpha F(E_{t_1}, Y(E_{t_1}), \alpha)(y)(gg^T)(t, t_2, y, \alpha)] \\
& + \int_{|z|<c} \int_0^1 \left[ \left( \partial_\alpha F(E_{t_1}, Y(E_{t_1}), \alpha)(y + \xi \tilde{h}(t, t_2, y, \alpha, z)) \right. \right. \\
& \quad \left. \left. - \partial_\alpha F(E_{t_1}, Y(E_{t_1}), \alpha)(y) \right) \tilde{h}(t, t_2, y, \alpha, z) \right] d\xi \nu(dz) \\
& + \int_{|z|<c} \left[ F(t_2, y + \tilde{h}(t, t_2, y, \alpha, z), \alpha_0) - F(t_2, y, \alpha_0) \right. \\
& \quad \left. - F_y(t_2, y, \alpha_0) \tilde{h}(t, t_2, y, \alpha, z) \right] \nu(dz),
\end{aligned}$$

where  $\alpha_t := \mathcal{L}(Y(t))$ ,  $t_1 \equiv t$ , and  $t_2 \equiv E_t$ .

**Proof: Part I.** Let  $f, g, \tilde{h}$  be bounded and define  $u(\alpha) = F(E, Y, \alpha)$  for fixed  $(E, Y) \in \mathbb{R}^+ \times \mathbb{R}^n$ . Then, we prove the Itô formula for  $u(\mathcal{L}(Y(t)))$ .

For any integer  $N \geq 1$ , define the empirical projection of  $u$  onto  $\mathbb{R}^n$  as:

$$u^N : (\mathbb{R}^n)^N \ni (Y^1(t), \dots, Y^N(t)) \mapsto u\left(\frac{1}{N} \sum_{l=1}^N \eta_{Y^l(t)}\right) = u(\bar{\alpha}_t^N), \quad (4.2)$$

where  $\eta_y$  denotes the unit mass at the point  $y \in \mathbb{R}^n$ . Then, Proposition 3.1 in [6] concludes that  $u^N$  is  $\mathbb{C}^2$  on  $\mathbb{R}^{n \times N}$  and

$$\partial_{y^i} u^N(Y^1, \dots, Y^N) = \frac{1}{N} \partial_\alpha u\left(\frac{1}{N} \sum_{l=1}^N \eta_{Y^l}\right)(Y^i) = \frac{1}{N} \partial_\alpha u(\bar{\alpha}_t^N)(Y^i), \quad (4.3)$$

$$\begin{aligned}
\partial_{y^i y^j}^2 u^N(Y^1, \dots, Y^N) &= \frac{1}{N} \partial_\nu \left[ \partial_\alpha u\left(\frac{1}{N} \sum_{l=1}^N \eta_{Y^l}\right) \right](Y^i) \delta_{i,j} + \frac{1}{N^2} \partial_\alpha^2 u\left(\frac{1}{N} \sum_{l=1}^N \eta_{Y^l}\right)(Y^i, Y^j) \\
&= \frac{1}{N} \partial_\nu \partial_\alpha u(\bar{\alpha}_t^N)(Y^i) \delta_{i,j} + \frac{1}{N^2} \partial_\alpha^2 u(\bar{\alpha}_t^N)(Y^i, Y^j),
\end{aligned} \quad (4.4)$$

where  $i, j = 1, 2, \dots, N$  and  $\delta_{i,j} = 1, i = j, \delta_{i,j} = 0, i \neq j$ . Moreover, let  $\{Y^\ell(t)\}_{t \in [0, T]}$  refer to a sequence of independent and identically distributed (i.i.d) copies of  $\{Y(t)\}_{t \in [0, T]}$ , i.e., for any  $\ell \geq 1$ ,

$$\begin{aligned}
dY^\ell(t) &= f(t, E_t, Y^\ell(t-), \mathcal{L}(Y^\ell(t-))) dE_t + g(t, E_t, Y^\ell(t-), \mathcal{L}(Y^\ell(t-))) dW_{E_t}^\ell \\
&\quad + \int_{|z|<c} \tilde{h}(t, E_t, Y^\ell(t-), \mathcal{L}(Y^\ell(t-)), z) \tilde{N}^\ell(dE_t, dz),
\end{aligned}$$

where  $W^\ell, N^\ell, \ell = 1, 2, \dots, N$  are mutually independent and have the same distributions to that of  $W, N$ , respectively. Applying Itô's formula [31] and taking the expectation, we have

$$\mathbb{E} u^N(Y^1(t), \dots, Y^N(t))$$

$$\begin{aligned}
&= \mathbb{E}u^N(Y^1(0), \dots, Y^N(0)) \\
&+ \sum_{\ell=1}^N \int_0^t \mathbb{E} \partial_{Y^\ell} u^N(Y^1(s), Y^2(s), \dots, Y^N(s)) f(s, E_s, Y^\ell(s-), \mathcal{L}(Y^\ell(s-))) dE_s \\
&+ \frac{1}{2} \sum_{\ell=1}^N \int_0^t \mathbb{E} \text{tr} \left[ (gg^T)(s, E_s, Y^\ell(s-), \mathcal{L}(Y^\ell(s-))) \partial_{Y^\ell}^2 u^N(Y^1(s), Y^2(s), \dots, Y^N(s)) \right] dE_s \\
&+ \int_0^t \int_{|z|<c} \mathbb{E} \left[ u^N(Y^1(s) + \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z), Y^2(s), \dots, Y^N(s)) \right. \\
&\quad \left. - u^N(Y^1(s), \dots, Y^N(s)) - \partial_{Y^1} u^N(Y^1(s), \dots, Y^N(s)) \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z) \right] \nu(dz) dE_s \\
&+ \dots \\
&+ \int_0^t \int_{|z|<c} \mathbb{E} \left[ u^N(Y^1(s), Y^2(s), \dots, Y^N(s) + \hbar(s, E_s, Y^N(s-), \mathcal{L}(Y^N(s-)), z)) \right. \\
&\quad \left. - u^N(Y^1(s), \dots, Y^N(s)) - \partial_{Y^N} u^N(Y^1(s), \dots, Y^N(s)) \hbar(s, E_s, Y^N(s-), \mathcal{L}(Y^N(s-)), z) \right] \nu(dy) dE_s \\
&= \mathbb{E}u^N(Y^1(0), \dots, Y^N(0)) \\
&+ N \int_0^t \mathbb{E} \partial_{Y^1} u^N(Y^1(s), Y^2(s), \dots, Y^N(s)) f(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-))) dE_s \\
&+ \frac{N}{2} \int_0^t \mathbb{E} \text{tr} \left[ (gg^T)(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-))) \partial_{Y^1}^2 u^N(Y^1(s), Y^2(s), \dots, Y^N(s)) \right] dE_s \\
&+ N \int_0^t \int_{|z|<c} \int_0^1 \mathbb{E} \left[ \left( \partial_{Y^1} u^N(Y^1(s) + \xi \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z), Y^2(s), \dots, Y^N(s)) \right. \right. \\
&\quad \left. \left. - \partial_{Y^1} u^N(Y^1(s), \dots, Y^N(s)) \right) \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z) \right] d\xi \nu(dz) dE_s,
\end{aligned}$$

where we have used the fact that the processes  $\{(Y^\ell(t))_{t \in [0, T]}\}_{\ell \in \{1, \dots, N\}}$  are i.i.d. in the second equality. Using Eqs (4.2)–(4.4), we obtain for any  $t \in J$

$$\begin{aligned}
&\mathbb{E}u^N(Y^1(t), \dots, Y^N(t)) \\
&= \mathbb{E}u^N(Y^1(0), \dots, Y^N(0)) \\
&\quad + \mathbb{E} \left[ \int_0^t \partial_\alpha u(\bar{\alpha}_s^N)(Y^1(s)) f(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-))) dE_s \right] \\
&\quad + \frac{1}{2} \mathbb{E} \left[ \int_0^t \text{tr} [\partial_\nu \partial_\alpha u(\bar{\alpha}_s^N)(Y^1(s)) (gg^T)(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)))] dE_s \right] \\
&\quad + \frac{1}{2N} \mathbb{E} \left[ \int_0^t \text{tr} [\partial_\alpha^2 u(\bar{\alpha}_s^N)(Y^1(s), Y^1(s)) (gg^T)(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)))] dE_s \right] \\
&\quad + \int_0^t \int_{|z|<c} \int_0^1 \mathbb{E} \left[ \left( \partial_\alpha u \left( \frac{1}{N} \eta_{Y^1(s) + \xi \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z)} + \frac{1}{N} \sum_{l=2}^N \eta_{Y^l(s)} \right) \right. \right. \\
&\quad \quad \left. \left. \circ (Y^1(s) + \xi \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z)) \right. \right. \\
&\quad \left. \left. - \partial_\alpha u(\bar{\alpha}_s^N)(Y^1(s)) \right) \hbar(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z) \right] d\xi \nu(dz) dE_s.
\end{aligned}$$

Taking  $N \rightarrow \infty$  and for any  $t \in J$ ,

$$\mathbb{P} \left[ \lim_{N \rightarrow \infty} \rho_2(\bar{\alpha}_t^N, \alpha_t)^2 = 0 \right] = 1.$$

Then, due to the continuity and boundedness of  $u, \partial_\alpha u, \partial_\nu \partial_\alpha u$  and the boundedness of  $\partial_\alpha^2 u, f, g$ , the dominated convergence theorem reads

$$\begin{aligned} & u(\mathcal{L}(Y(t))) \\ &= u(\mathcal{L}(Y(0))) \\ &+ \mathbb{E} \left[ \int_0^t \partial_\alpha u(\mathcal{L}(Y(s)))(Y^1(s)) f(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-))) dE_s \right] \\ &+ \frac{1}{2} \mathbb{E} \left[ \int_0^t \text{tr}[\partial_\nu \partial_\alpha u(\mathcal{L}(Y(s)))(Y^1(s))(gg^T)(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)))] dE_s \right] \\ &+ \int_0^t \int_{|z|<c} \int_0^1 \mathbb{E} \left[ \left( \partial_\alpha u(\mathcal{L}(Y(s)))(Y^1(s)) + \xi \bar{h}(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z) \right. \right. \\ &\quad \left. \left. - \partial_\alpha u(\mathcal{L}(Y(s)))(Y^1(s)) \right) \bar{h}(s, E_s, Y^1(s-), \mathcal{L}(Y^1(s-)), z) \right] d\xi \nu(dz) dE_s. \end{aligned} \quad (4.5)$$

**Part II.** Assuming that Conditions 3.1–3.3 hold, the Itô formula for  $u(\mathcal{L}(Y(t)))$  will be derived.

For every  $n \in \mathbb{N}$ , we take a smooth function  $\psi_n : \mathbb{R}^n \mapsto \mathbb{R}^n$  satisfying:

$$\begin{cases} \psi_n(y) = y, & |y| \leq n, \\ \psi_n(y) = 0, & |y| > 2n, \end{cases}$$

such that

$$|\psi_n(y)| \leq C_1, \quad |\partial \psi_n(y)| \leq C_1, \quad (4.6)$$

where the constant  $C_1 > 0$  is independent of  $n$ . Set

$$f^{(n)}(t_1, t_2, y, \alpha) := f(t_1, t_2, \psi_n(y), \alpha), \quad g^{(n)}(t_1, t_2, y, \alpha) := g(t_1, t_2, \psi_n(y), \alpha),$$

$$\bar{h}^{(n)}(t_1, t_2, y, \alpha, z) := \bar{h}(t_1, t_2, \psi_n(y), \alpha, z).$$

Allowing  $n \rightarrow \infty$ ,

$$f^{(n)}(t_1, t_2, y, \alpha) \rightarrow f(t_1, t_2, y, \alpha), \quad g^{(n)}(t_1, t_2, y, \alpha) \rightarrow g(t_1, t_2, y, \alpha),$$

$$\bar{h}^{(n)}(t_1, t_2, y, \alpha, z) \rightarrow \bar{h}(t_1, t_2, y, \alpha, z).$$

Furthermore, by Condition 3.2, we notice that  $f^{(n)}, g^{(n)}$ , and  $\bar{h}^{(n)}$  are bounded, and all  $f^{(n)}, g^{(n)}, \bar{h}^{(n)}$  obey Conditions 3.1 and 3.3. Hence, the following equation

$$dY^{(n)}(t) = f^{(n)}(t, E_t, Y^{(n)}(t-), \mathcal{L}(Y^{(n)}(t-))) dE_t$$

$$\begin{aligned}
& +g^{(n)}(t, E_t, Y^{(n)}(t-), \mathcal{L}(Y^{(n)}(t-)))dW_{E_t} \\
& + \int_{|z|<c} \tilde{h}^{(n)}(t, E_t, Y^{(n)}(t-), \mathcal{L}(Y^{(n)}(t-)), z)\tilde{N}(dE_t, dz)
\end{aligned}$$

has a unique solution  $Y^{(n)}(t)$  with  $Y^{(n)}(0) = Y(0)$ . Then, for any  $t \in J$  and according to **Part I**, we conclude that

$$\begin{aligned}
& u(\mathcal{L}(Y^{(n)}(t))) \\
& = u(\mathcal{L}(Y^{(n)}(0))) \\
& + \mathbb{E} \left[ \int_0^t \partial_\alpha u(\mathcal{L}(Y^{(n)}(s)))(Y^{(n)}(s))f^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-)))dE_s \right] \\
& + \frac{1}{2} \mathbb{E} \left[ \int_0^t tr[\partial_\alpha u(\mathcal{L}(Y^{(n)}(s)))(Y^{(n)}(s))(g^{(n)}g^{(n)T})(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-)))dE_s \right] \\
& + \int_0^t \int_{|z|<c} \int_0^1 \mathbb{E} \left[ \left( \partial_\alpha u(\mathcal{L}(Y^{(n)}(s)))(Y^{(n)}(s)) + \xi \tilde{h}^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-)), z) \right. \right. \\
& \quad \left. \left. - \partial_\alpha u(\mathcal{L}(Y^{(n)}(s)))(Y^{(n)}(s)) \right) \tilde{h}^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-)), z) \right] d\xi v(dz)dE_s. \tag{4.7}
\end{aligned}$$

Taking the expectation, using Hölder's inequality, the Burkholder-Davis-Gundy inequality [12], Doob's martingale inequality, and Itô's isometry (from [39]), we have

$$\begin{aligned}
& \mathbb{E}_W |Y^{(n)}(t) - Y(t)|^2 \\
& \leq 3E_T \mathbb{E}_W \int_0^t |f^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-))) - f(s, E_s, Y(s-), \mathcal{L}(Y(s-)))|^2 dE_s \\
& \quad + 3b_2 \mathbb{E}_W \int_0^t \|g^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-))) - g(s, E_s, Y(s-), \mathcal{L}(Y(s-)))\|^2 dE_s \\
& \quad + 12 \mathbb{E}_W \int_0^t \int_{|z|<c} |\tilde{h}^{(n)}(s, E_s, Y^{(n)}(s-), \mathcal{L}(Y^{(n)}(s-)), z) \\
& \quad - \tilde{h}(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z)|^2 v(dz)dE_s.
\end{aligned}$$

By Condition 3.1 and the Jensen inequality, we have

$$\begin{aligned}
\mathbb{E}_W |Y^{(n)}(t) - Y(t)|^2 & \leq 3[E_T + b_2 + 4] \mathbb{E}_W \int_0^t L \Gamma \left( |\psi_n(Y^{(n)}(s-)) - Y(s-)|^2 \right. \\
& \quad \left. + \rho_2(\mathcal{L}(Y^{(n)}(s-)), \mathcal{L}(Y(s-)))^2 \right) dE_s \\
& \leq 3L[E_T + b_2 + 4] \int_0^t \Gamma \left( \mathbb{E}_W |\psi_n(Y^{(n)}(s-)) - Y(s-)|^2 \right. \\
& \quad \left. + \mathbb{E}_W |Y^{(n)}(s-) - Y(s-)|^2 \right) dE_s.
\end{aligned}$$

By Eq (4.6), we have

$$\begin{aligned}
\mathbb{E}_W |\psi_n(Y^{(n)}(s)) - Y(s)|^2 & \leq 2\mathbb{E}_W |\psi_n(Y^{(n)}(s)) - \psi_n(Y(s))|^2 + 2\mathbb{E}_W |\psi_n(Y(s)) - Y(s)|^2 \\
& \leq 2C_1 \mathbb{E}_W |Y^{(n)}(s) - Y(s)|^2 + 2\mathbb{E}_W |\psi_n(Y(s)) - Y(s)|^2,
\end{aligned}$$

and furthermore,

$$\begin{aligned} \mathbb{E}_W |Y^{(n)}(t) - Y(t)|^2 &\leq 3L[E_T + b_2 + 4] \int_0^t \Gamma \left( (2C_1 + 1) \mathbb{E}_W |Y^{(n)}(s-) - Y(s-)|^2 \right. \\ &\quad \left. + 2\mathbb{E}_W |\psi_n(Y(s-)) - Y(s-)|^2 \right) dE_s. \end{aligned}$$

Therefore,

$$\begin{aligned} &(2C_1 + 1) \mathbb{E}_W |Y^{(n)}(t) - Y(t)|^2 + 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 \\ &\leq 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 + 3L[E_T + b_2 + 4](2C_1 + 1) \\ &\quad \times \int_0^t \Gamma \left( (2C_1 + 1) \mathbb{E}_W |Y^{(n)}(s-) - Y(s-)|^2 + 2\mathbb{E}_W |\psi_n(Y(s-)) - Y(s-)|^2 \right) dE_s. \end{aligned}$$

Putting  $G(t) = \int_0^t \frac{ds}{\Gamma(s)}$ , it is concluded from Lemma 3.1 that

$$\begin{aligned} &(2C_1 + 1) \mathbb{E}_W |Y^{(n)}(t) - Y(t)|^2 + 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 \\ &\leq G^{-1} \left[ G \left( 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 \right) + 3L[E_T + b_2 + 4](2C_1 + 1)E_T \right]. \end{aligned}$$

Using the fact that  $\lim_{n \rightarrow \infty} \psi_n(y) = y$  for all  $y \in \mathbb{R}^n$ , then  $\lim_{n \rightarrow \infty} \mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 = 0$ . Referring to the hypothesis  $\int_{0^+} \frac{ds}{\Gamma(s)} = \infty$ , it yields that

$$G \left( 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 \right) + 3[E_T + b_2 + 4]L(2C_1 + 1)E_T \longrightarrow -\infty, \quad n \longrightarrow \infty.$$

On the other side, since  $G$  is a strictly increasing function, it is obtained that  $G$  has an inverse function that is strictly increasing, and  $G^{-1}(-\infty) = 0$ , i.e.,

$$G^{-1} \left[ G \left( 2\mathbb{E}_W |\psi_n(Y(t)) - Y(t)|^2 \right) + 3L[E_T + b_2 + 4](2C_1 + 1)E_T \right] \longrightarrow 0, \quad n \longrightarrow \infty.$$

We concluded that

$$\lim_{n \rightarrow \infty} \mathbb{E} |Y^{(n)}(t) - Y(t)|^2 = 0.$$

Moreover, we can get

$$\lim_{n \rightarrow \infty} \mathbb{E} |\psi_n(Y^{(n)}(t)) - Y(t)|^2 = 0$$

and

$$\lim_{n \rightarrow \infty} \rho_2 \left( \mathcal{L}(Y^{(n)}(t)), \mathcal{L}(Y(t)) \right)^2 \leq \lim_{n \rightarrow \infty} \mathbb{E} |Y^{(n)}(t) - Y(t)|^2 = 0.$$

By taking the limit on both sides of Eq (4.7), the dominated convergence theorem concludes (4.5).

**Part III.** We prove Itô's formula (4.1). By Itô's formula ([31, Lemma 3.1]) and Eq (4.5), it can be concluded that

$$\begin{aligned}
& F(E_t, Y(t), \alpha_t) - F(0, y_0, \alpha_0) \\
&= F(E_t, Y(t), \alpha_t) - F(E_t, Y(t), \alpha_0) + F(E_t, Y(t), \alpha_0) - F(0, y_0, \alpha_0) \\
&= \int_0^t \partial_{t_2} F(E_s, Y(s-), \alpha_0) dE_s + \int_0^t F_y(E_s, Y(s-), \alpha_0) f(s, E_s, Y(s-), \mathcal{L}(Y(s-))) dE_s \\
&\quad + \frac{1}{2} \int_0^t \text{tr} \left[ F_{yy}(E_s, Y(s-), \alpha_0) (gg^T)(s, E_s, Y(s-), \mathcal{L}(Y(s-))) \right] dE_s \\
&\quad + \int_0^t \int_{|z| < c} \left[ F(E_s, Y(s-) + \hbar(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z), \alpha_0) - F(E_s, Y(s-), \alpha_0) \right. \\
&\quad \left. - F_y(E_s, Y(s-), \alpha_0) \hbar(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z) \right] \nu(dz) dE_s \\
&\quad + \int_0^t \int_{|z| < c} \left[ F(E_s, Y(s-) + \hbar(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z), \alpha_0) \right. \\
&\quad \left. - F(E_s, Y(s-), \alpha_0) \right] \tilde{N}(dz, dE_s) \\
&\quad + \int_0^t F_y(E_s, Y(s-), \alpha_0) g(s, E_s, Y(s-), \mathcal{L}(Y(s-))) dW_{E_s} \\
&\quad + \mathbb{E} \left[ \int_0^t \partial_\alpha F(E_t, Y(t-), \alpha_s) (Y(s)) f(s, E_s, Y(s-), \mathcal{L}(Y(s-))) dE_s \right] \\
&\quad + \frac{1}{2} \mathbb{E} \left[ \int_0^t \text{tr} [\partial_v \partial_\alpha F(E_t, Y(t-), \alpha_s) (Y(s)) (gg^T)(s, E_s, Y(s-), \mathcal{L}(Y(s-)))] dE_s \right] \\
&\quad + \int_0^t \int_{|z| < c} \int_0^1 \mathbb{E} \left[ \left( \partial_\alpha F(E_t, Y(t-), \alpha_s) (Y(s)) + \xi \hbar(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z) \right) \right. \\
&\quad \left. - \partial_\alpha F(E_t, Y(t-), \alpha_s) (Y(s)) \right] \hbar(s, E_s, Y(s-), \mathcal{L}(Y(s-)), z) d\xi \nu(dz) dE_s.
\end{aligned}$$

This completes the proof.  $\square$

**Remark 4.1.** Let  $F$  be independent of the measure  $\alpha$ . Then  $F \in \wp_b(\mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$  is simplified to  $\mathbb{C}^{1,2}(\mathbb{R}^+ \times \mathbb{R}^n)$  and the relevant Itô formula reduces to the standard Itô formula introduced by ([31, Lemma 3.1]).

## 5. Stability results

In this section, we study the stability of the solutions of Eq (1.2) based on the Lyapunov method. We establish some sufficient criteria to derive the  $p$ -th moment asymptotic stability, stability in probability, asymptotic stability in probability, and globally asymptotic stability in probability for the trivial solution, generalizing some results obtained in [31, 40]. For this purpose, let the initial value  $Y(0) = y_0 \in \mathbb{R}^n$  be deterministic constant. Furthermore, let  $\mathcal{H}$  be the set of all non-decreasing functions  $\phi : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  such that  $\phi(s) > 0$ ,  $s > 0$ . Besides, for any  $h > 0$ , define  $\mathcal{A}_h := \{y \in \mathbb{R}^n : |y| < h\}$  and  $\bar{\mathcal{A}}_h := \{y \in \mathbb{R}^n : |y| \leq h\}$ .

**Definition 5.1.** ([31, 45]) The trivial solution of Eq (1.2) is called  $p$ -th moment asymptotically stable if for all  $t \geq 0$ , there exist a function  $\nu(t) : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  decaying to 0 as  $t \rightarrow \infty$  and  $C > 0$  such that

$$\mathbb{E}[|Y(t)|^p] < C|y_0|^p \nu(t),$$

where  $p > 0$  and  $y_0 \in \mathbb{R}^n$  is the initial value.

**Theorem 5.1.** Let  $p, a_1, a_2, a_3$  be positive constants. If  $F(t_2, y, \alpha) \in \wp_{b,+}(\mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$  satisfies

- (1)  $F(t_2, 0, \delta_0) = 0$ ,
- (2)  $a_1|y|^{2p} \leq F(t_2, y, \alpha) \leq a_2|y|^{2p}$ ,
- (3)  $\mathbb{L}_{f,g,\tilde{h}}F(t_2, y, \alpha) + a_3F(t_2, y, \alpha_0) \leq 0$ ,

for all  $(t_2, y, \alpha) \in \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n)$ , then the trivial solution of Eq (1.2) is  $p$ -th moment asymptotically stable.

*Proof.* Fix any  $y_0 \neq 0$  in  $\mathbb{R}^n$ . For each  $m \geq |y_0|$ , define

$$\tau_m = \inf\{t \geq 0 : |Y(t)| \geq m\},$$

$$\mathcal{U}_k := k \wedge \inf\left\{t \geq 0 : \left| \int_0^{\tau_m \wedge t} F_y(E_s, Y(s), \alpha_0)g(s, E_s, Y(s), \mathcal{L}(Y(s-)))dW_{E_s} \right| \geq k\right\}, \quad (5.1)$$

$$\begin{aligned} \mathcal{W}_k := k \wedge \inf\left\{t \geq 0 : \left| \int_0^{\tau_m \wedge t} \int_{|z|<c} [F(E_s, Y(s) + \tilde{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \right. \right. \\ \left. \left. - F(E_s, Y(s), \alpha_0)]\tilde{N}(dz, dE_s) \right| \geq k\right\}, \end{aligned} \quad (5.2)$$

for  $k = 1, 2, \dots$ . It is obvious that  $\mathcal{U}_k \rightarrow \infty$  and  $\mathcal{W}_k \rightarrow \infty$  as  $k \rightarrow \infty$ . Applying Lemma 4.1 for the function  $e^{a_3 t_2} F(t_2, y, \alpha)$ , we obtain

$$\begin{aligned} & e^{a_3 E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}} F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}) \\ &= F(0, y_0, \alpha_0) + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} [a_3 F(E_s, Y(s), \alpha_0) + \partial_{t_2} F(E_s, Y(s), \alpha_0)] dE_s \\ &+ \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} F_y(E_s, Y(s), \alpha_0) f(s, E_s, Y(s), \mathcal{L}(Y(s-))) dE_s \\ &+ \frac{1}{2} \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} \text{tr}[F_{yy}(E_s, Y(s), \alpha_0)(gg^T)(s, E_s, Y(s), \mathcal{L}(Y(s-)))] dE_s \\ &+ \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z|<c} e^{a_3 E_s} [F(E_s, Y(s) + \tilde{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \\ &- F(E_s, Y(s), \alpha_0) - F_y(E_s, Y(s), \alpha_0)\tilde{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z)] \nu(dz) dE_s \\ &+ \mathbb{E} \left[ \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} \left( \partial_\alpha F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_s)(Y(s)) \right. \right. \\ &\left. \left. \times f(s, E_s, Y(s), \mathcal{L}(Y(s))) + \frac{1}{2} \text{tr}[\partial_\nu \partial_\alpha F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_s)(Y(s)) \right] \right) \end{aligned}$$

$$\begin{aligned}
& \times (gg^T)(s, E_s, Y(s), \mathcal{L}(Y(s))) \Big] dE_s \\
& + \mathbb{E} \left[ \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} \int_0^1 e^{a_3 E_s} \left( \partial_\alpha F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_s)(Y(s) \right. \right. \\
& \left. \left. + \xi \hbar(s, E_s, Y(s), \mathcal{L}(Y(s)), z) - \partial_\alpha F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_s)(Y(s)) \right) \right. \\
& \left. \times \hbar(s, E_s, Y(s), \mathcal{L}(Y(s)), z) d\xi v(dz) \right] dE_s \\
& + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s} \\
& + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} e^{a_3 E_s} [F(E_s, Y(s) + \hbar(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \\
& - F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s) \\
& = F(0, y_0, \alpha_0) + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} \left[ a_3 F(E_s, Y(s), \alpha_0) + \mathbb{L}_{f, g, \hbar} F(E_s, Y(s), \alpha_s) \right] dE_s \\
& + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s} \\
& + \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} e^{a_3 E_s} [F(E_s, Y(s) + \hbar(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \\
& - F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s). \tag{5.3}
\end{aligned}$$

We have by Kuo [17] and Magdziarz [23] that

$$\int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} e^{a_3 E_s} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s}$$

and

$$\begin{aligned}
& \int_0^{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} e^{a_3 E_s} [F(E_s, Y(s) + \hbar(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \\
& - F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s)
\end{aligned}$$

are martingales with mean 0. Taking the expectation on both sides of Eq (5.3) and using assumption (3), we have that

$$\mathbb{E} \left( e^{a_3 E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}} F(E_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_{t \wedge \tau_m \wedge \mathcal{U}_k \wedge \mathcal{W}_k}) \right) \leq F(0, y_0, \alpha_0).$$

Allowing  $k, m \rightarrow \infty$ , we get

$$\mathbb{E} \left( e^{a_3 E_t} F(E_t, Y(t), \alpha_t) \right) \leq F(0, y_0, \alpha_0).$$

Then assumption (2) implies

$$a_1 \mathbb{E} \left( e^{a_3 E_t} |Y(t)|^{2p} \right) \leq \mathbb{E} \left( e^{a_3 E_t} F(E_t, Y(t), \alpha_t) \right) \leq F(0, y_0, \alpha_0) \leq a_2 |y_0|^{2p},$$

which gives

$$\mathbb{E}(e^{a_3 E_t} |Y(t)|^{2p}) \leq \frac{a_2}{a_1} |y_0|^{2p}.$$

Using the reverse Hölder inequality  $\mathbb{E}|\mathcal{X}\mathcal{Y}| \geq (\mathbb{E}|\mathcal{X}|^{\frac{1}{p}})^p (\mathbb{E}|\mathcal{Y}|^{\frac{-1}{p-1}})^{-(p-1)}$  when  $p = 2$ ,  $\mathcal{X} = |Y(t)|^{2p}$ , and  $\mathcal{Y} = e^{a_3 E_t}$ , we have

$$(\mathbb{E}|Y(t)|^p)^2 (\mathbb{E}(e^{-a_3 E_t}))^{-1} \leq \frac{a_2}{a_1} |y_0|^{2p}.$$

Additionally, the inverse  $\beta$ -stable subordinator  $E_t$  takes the Laplace transform  $\mathbb{E}(e^{-a_3 E_t}) = E_\beta(-a_3 t^\beta)$ , where the function  $E_\beta(t) = \sum_{n=0}^{\infty} \frac{t^n}{\Gamma(n\beta+1)}$  is the Mittag-Leffler function. Then

$$\mathbb{E}|Y(t)|^p \leq \sqrt{\frac{a_2}{a_1}} |y_0|^p \sqrt{E_\beta(-a_3 t^\beta)},$$

with  $E_\beta(-a_3 t^\beta) \rightarrow 0$  as  $t \rightarrow \infty$  [24]. Therefore, the trivial solution of Eq (1.2) is  $p$ -th moment asymptotically stable and the proof is complete.  $\square$

**Remark 5.1.** Putting  $\hbar \equiv 0$  in Eq (1.2), our results in Theorem 5.1 are new for MFSDEs driven by time-changed Brownian motion studied by [21, 40].

**Definition 5.2.** ([31, 45]) If for every pair of  $\varsigma > 0$  and  $0 < \varepsilon < 1$ , there exists  $0 < \delta = \delta(\varepsilon, \varsigma)$  such that when  $|y_0| < \delta$ ,

$$\mathbb{P}\{|Y(t, y_0)| < \varsigma \text{ for } \forall t \geq 0\} \geq 1 - \varepsilon,$$

then the trivial solution of Eq (1.2) is called stable in probability.

**Theorem 5.2.** Suppose for all  $(t_2, y, \alpha) \in \mathbb{R}^+ \times \mathcal{A}_h \times \mathcal{M}_2(\mathbb{R}^n)$ , there exists a function  $F(t_2, y, \alpha) \in \wp_{b,+}(\mathbb{R}^+ \times \mathcal{A}_h \times \mathcal{M}_2(\mathbb{R}^n))$  with  $h > 2c$  and  $\phi \in \mathcal{H}$  such that

- (1)  $F(t_2, 0, \delta_0) = 0$ ,
- (2)  $\phi(|y|) \leq F(t_2, y, \alpha)$ ,
- (3)  $\mathbb{L}_{f,g,\hbar} F(t_2, y, \alpha) \leq 0$ ,

then the trivial solution of Eq (1.2) is stable in probability.

*Proof.* Select  $\varepsilon \in (0, 1)$  and  $\varsigma \in (0, h)$  arbitrarily. Since  $F(t_2, y, \alpha)$  is continuous on  $\mathbb{R}^+ \times \mathcal{A}_h \times \mathcal{M}_2(\mathbb{R}^n)$  and  $F(t_2, 0, \delta_0) = 0$ ,  $\exists$  a  $\Theta = \Theta(\varepsilon, \varsigma) > 0$  such that

$$\frac{1}{\varepsilon} \sup_{y_0 \in \mathcal{A}_\Theta} F(0, y_0, \alpha_0) \leq \phi(\varsigma). \quad (5.4)$$

Then by assumption (2) and (5.4), we have  $\Theta < \varsigma$ . Define the following stopping times:

$$\tau_\varsigma := \inf\{t \geq 0 : |Y(t, y_0)| \geq \varsigma\}, \quad (5.5)$$

$$\mathcal{U}_k := k \wedge \inf\left\{t \geq 0 : \left| \int_0^{\tau_\varsigma \wedge t} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s} \right| \geq k\right\}, \quad (5.6)$$

and

$$\mathcal{W}_k := k \wedge \inf \left\{ t \geq 0 : \left| \int_0^{\tau_\zeta \wedge t} \int_{|z| < c} \left[ F(E_s, Y(s) + \mathfrak{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) - F(E_s, Y(s), \alpha_0) \right] \tilde{N}(dz, dE_s) \right| \geq k \right\}, \quad (5.7)$$

for all  $k = 1, 2, \dots$ , where the initial value  $y_0 \in \mathcal{A}_\Theta$  is fixed. It is clear that  $\mathcal{U}_k \rightarrow \infty$  and  $\mathcal{W}_k \rightarrow \infty$  as  $k \rightarrow \infty$ . Applying Lemma 4.1 to Eq (1.2), we have

$$\begin{aligned} & F(E_{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k}) \\ &= F(0, y_0, \alpha_0) + \int_0^{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \mathbb{L}_{f, g, \mathfrak{h}} F(E_s, Y(s), \alpha_s) dE_s \\ &+ \int_0^{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s} \\ &+ \int_0^{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} [F(E_s, Y(s) + \mathfrak{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \\ &- F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s). \end{aligned} \quad (5.8)$$

Owing to Kuo [17] and Magdziarz [23], the following:

$$\int_0^{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s}$$

and

$$\int_0^{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k} \int_{|z| < c} [F(E_s, Y(s) + \mathfrak{h}(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) - F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s)$$

are square integrable martingales with respect to  $\mathcal{G}_t = \mathfrak{F}_{E_t}$  having mean 0. Using assumption (3) and taking the expectation on all sides of (5.8), we get

$$\mathbb{E}F(E_{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k}, Y(\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_{\tau_\zeta \wedge \mathcal{U}_k \wedge \mathcal{W}_k}) \leq F(0, y_0, \alpha_0).$$

Then by letting  $k \rightarrow \infty$ ,

$$\mathbb{E}F(E_{\tau_\zeta}, Y(\tau_\zeta), \alpha_{\tau_\zeta}) \leq F(0, y_0, \alpha_0).$$

As  $F(t_2, y, \alpha) > 0$ ,

$$\mathbb{E}\{F(E_{\tau_\zeta}, Y(\tau_\zeta), \alpha_{\tau_\zeta})1_{\{\tau_\zeta < \infty\}}\} \leq \mathbb{E}F(E_{\tau_\zeta}, Y(\tau_\zeta), \alpha_{\tau_\zeta}) \leq F(0, y_0, \alpha_0). \quad (5.9)$$

Assumption (2) reads

$$\mathbb{E}\{\phi(|Y(\tau_\zeta)|)1_{\{\tau_\zeta < \infty\}}\} \leq \mathbb{E}\{F(E_{\tau_\zeta}, Y(\tau_\zeta), \alpha_{\tau_\zeta})1_{\{\tau_\zeta < \infty\}}\}. \quad (5.10)$$

Using Eqs (5.5), and (5.10) and as the function  $\phi$  is non-decreasing, we obtain

$$|Y(\tau_\varsigma)| \geq \varsigma$$

and

$$\phi(\varsigma)\mathbb{P}\{\tau_\varsigma < \infty\} \leq \mathbb{E}\{F(E_{\tau_\varsigma}, Y(\tau_\varsigma), \alpha_{\tau_\varsigma})1_{\{\tau_\varsigma < \infty\}}\}. \quad (5.11)$$

Combining (5.4), (5.9), and (5.11), we get  $\mathbb{P}\{\tau_\varsigma < \infty\} \leq \varepsilon$ , which indicates  $\mathbb{P}\{\tau_\varsigma = \infty\} \geq 1 - \varepsilon$ , i.e.,

$$\mathbb{P}\{|Y(t, y_0)| < \varsigma \text{ for all } t \geq 0\} \geq 1 - \varepsilon.$$

Hence  $Y(t, y_0)$  is stable in probability and the proof is finished.  $\square$

**Definition 5.3.** ([31, 45]) If for every  $\varepsilon \in (0, 1)$ , there exists  $0 < \delta_0 = \delta_0(\varepsilon)$  such that when  $|y_0| < \delta_0$ ,

$$\mathbb{P}\{\lim_{t \rightarrow \infty} Y(t, y_0) = 0\} \geq 1 - \varepsilon,$$

then the trivial solution of (1.2) is called asymptotically stable in probability provided that it is stable in probability.

**Theorem 5.3.** Suppose for all  $(t_2, y, \alpha) \in \mathbb{R}^+ \times \mathcal{A}_h \times \mathcal{M}_2(\mathbb{R}^n)$  there exists a function  $F(t_2, y, \alpha) \in \wp_{b,+}(\mathbb{R}^+ \times \mathcal{A}_h \times \mathcal{M}_2(\mathbb{R}^n))$  with  $h > 2c$  and  $\phi \in \mathcal{H}$  such that

- (1)  $F(t_2, 0, \delta_0) = 0$ ,
- (2)  $\phi(|y|) \leq F(t_2, y, \alpha)$ ,
- (3) for any  $\mu \in (0, h)$ ,  $y \in \mathcal{A}_h - \bar{\mathcal{A}}_\mu$ ,  $\mathbb{L}_{f,g,h}F(t_2, y, \alpha) \leq -\gamma(\mu)$  a.s.,

where  $\gamma(\mu) \geq 0$  and  $\gamma(\mu) \neq 0$  when  $\mu \neq 0$ . Then the trivial solution of (1.2) is asymptotically stable in probability.

*Proof.* Depending on Theorem 5.2 and for any  $0 < \varepsilon < 1$ , there exists a positive  $\Theta = \Theta(\varepsilon)$  such that

$$\mathbb{P}\{|Y(t, y_0)| < h \text{ for } \forall t \geq 0\} \geq 1 - \frac{\varepsilon}{5}, \quad (5.12)$$

where  $y_0 \in \mathcal{A}_\Theta$ . Let  $0 < \mu < \varpi < |y_0|$  be arbitrary and define

$$\tau_h := \inf\{t \geq 0 : |Y(t, y_0)| \geq h\}, \quad \tau_\mu := \inf\{t \geq 0 : |Y(t, y_0)| \leq \mu\},$$

$$\begin{aligned} \mathcal{U}_k &:= k \wedge \inf\left\{t \geq 0 : \left| \int_0^{\tau_h \wedge \tau_\mu \wedge t} F_y(E_s, Y(s), \alpha_0)g(s, E_s, Y(s), \mathcal{L}(Y(s-)))dW_{E_s} \right| \geq k\right\}, \\ \mathcal{W}_k &:= k \wedge \inf\left\{t \geq 0 : \left| \int_0^{\tau_h \wedge \tau_\mu \wedge t} \int_{|z| < c} [F(E_s, Y(s) + \hbar(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \right. \right. \\ &\quad \left. \left. - F(E_s, Y(s), \alpha_0)]\tilde{N}(dz, dE_s) \right| \geq k\right\}, \end{aligned} \quad (5.13)$$

for  $k = 1, 2, \dots$ , where  $y_0 \in \mathcal{A}_\Theta$  is fixed. Clearly,  $\mathcal{U}_k, \mathcal{W}_k \rightarrow \infty$  as  $k \rightarrow \infty$ . Applying Lemma 4.1 and assumption (3) to (1.2), we have

$$\begin{aligned} 0 &\leq \mathbb{E}F(E_{t \wedge \tau_h \wedge \tau_\mu} \mathcal{U}_k \wedge \mathcal{W}_k, Y(t \wedge \tau_h \wedge \tau_\mu \wedge \mathcal{U}_k \wedge \mathcal{W}_k), \alpha_{t \wedge \tau_h \wedge \tau_\mu} \mathcal{U}_k \wedge \mathcal{W}_k) \\ &\leq F(0, y_0, \alpha_0) - \gamma(\mu) \mathbb{E}E_{t \wedge \tau_h \wedge \tau_\mu} \mathcal{U}_k \wedge \mathcal{W}_k. \end{aligned}$$

Putting  $t, k \rightarrow \infty$ , we get

$$\gamma(\mu) \mathbb{E}E_{\tau_h \wedge \tau_\mu} \leq F(0, y_0, \alpha_0).$$

Because  $E_t \rightarrow \infty$  as  $t \rightarrow \infty$  and  $\gamma(\mu) \neq 0$ ,

$$\mathbb{P}\{\tau_h \wedge \tau_\mu < \infty\} = 1. \quad (5.14)$$

From (5.12) and (5.13), it follows that  $\mathbb{P}(\tau_h < \infty) \leq \frac{\varepsilon}{5}$ . Therefore,

$$1 = \mathbb{P}\{\tau_h \wedge \tau_\mu < \infty\} \leq \mathbb{P}\{\tau_h < \infty\} + \mathbb{P}\{\tau_\mu < \infty\} \leq \mathbb{P}\{\tau_\mu < \infty\} + \frac{\varepsilon}{5}.$$

Hence,  $\mathbb{P}\{\tau_\mu < \infty\} \geq 1 - \frac{\varepsilon}{5}$ . Also, there exists a positive constant  $\bar{\Theta} = \bar{\Theta}(\mu)$  such that

$$\mathbb{P}\{\tau_\mu < \bar{\Theta}\} \geq 1 - \frac{2\varepsilon}{5}.$$

Additionally,

$$\begin{aligned} \mathbb{P}\{\tau_\mu < \tau_h \wedge \bar{\Theta}\} &\geq \mathbb{P}(\{\tau_\mu < \bar{\Theta}\} \cap \{\tau_h = \infty\}) \\ &= \mathbb{P}(\tau_\mu < \bar{\Theta}) - \mathbb{P}(\{\tau_\mu < \bar{\Theta}\} \cap \{\tau_h < \infty\}) \\ &\geq \mathbb{P}(\tau_\mu < \bar{\Theta}) - \mathbb{P}(\tau_h < \infty) \geq 1 - \frac{2\varepsilon}{5} - \frac{\varepsilon}{5} = 1 - \frac{3\varepsilon}{5}. \end{aligned} \quad (5.15)$$

We define the stopping times

$$\varrho := \begin{cases} \tau_\mu, & \text{if } \tau_\mu < \tau_h \wedge \bar{\Theta}, \\ \infty, & \text{otherwise,} \end{cases} \quad (5.16)$$

$$\begin{aligned} \tau_\varpi &:= \inf\{t \geq \varrho : |Y(t, y_0)| \geq \varpi\}, \\ M_\ell &:= \inf\left\{t \geq \varrho : \left| \int_\varrho^{\tau_\varpi \wedge t} F_y(E_s, Y(s), \alpha_0) g(s, E_s, Y(s), \mathcal{L}(Y(s-))) dW_{E_s} \right| \geq \ell\right\}, \\ R_\ell &:= \inf\left\{t \geq \varrho : \left| \int_\varrho^{\tau_\varpi \wedge t} \int_{|z| < c} [F(E_s, Y(s) + \hbar(s, E_s, Y(s), \mathcal{L}(Y(s-)), z), \alpha_0) \right. \right. \\ &\quad \left. \left. - F(E_s, Y(s), \alpha_0)] \tilde{N}(dz, dE_s) \right| \geq \ell\right\}, \end{aligned} \quad (5.17)$$

for  $\ell = 1, 2, \dots$ , where  $M_\ell, R_\ell \rightarrow \infty$  as  $\ell \rightarrow \infty$ . Again, using Lemma 4.1 and assumption (3), we have

$$\begin{aligned} &\mathbb{E}F(E_{t \wedge \tau_h \wedge \tau_\varpi \wedge M_\ell \wedge R_\ell}, Y(t \wedge \tau_h \wedge \tau_\varpi \wedge M_\ell \wedge R_\ell), \alpha_{t \wedge \tau_h \wedge \tau_\varpi \wedge M_\ell \wedge R_\ell}) \\ &\leq \mathbb{E}F(E_{t \wedge \tau_h \wedge \varrho}, Y(t \wedge \tau_h \wedge \varrho), \alpha_{t \wedge \tau_h \wedge \varrho}) \end{aligned}$$

for all  $\ell = 1, 2, \dots$  and  $t \in \mathbb{R}^+$ . Letting  $\ell \rightarrow \infty$  followed by  $t \rightarrow \infty$ ,

$$\mathbb{E}F(E_{\tau_{\varpi} \wedge \tau_h}, Y(\tau_{\varpi} \wedge \tau_h), \alpha_{\tau_{\varpi} \wedge \tau_h}) \leq \mathbb{E}F(E_{\varrho \wedge \tau_h}, Y(\varrho \wedge \tau_h), \alpha_{\varrho \wedge \tau_h}). \quad (5.18)$$

Combining Eqs (5.16), (5.17), and (5.18),

$$\mathbb{E}\{F(E_{\tau_{\varpi} \wedge \tau_h}, Y(\tau_{\varpi} \wedge \tau_h), \alpha_{\tau_{\varpi} \wedge \tau_h})1_{\{\tau_{\varpi} < \infty\}}\} \leq \mathbb{E}\{F(E_{\varrho \wedge \tau_h}, Y(\varrho \wedge \tau_h), \alpha_{\varrho \wedge \tau_h})1_{\{\varrho < \infty\}}\}, \quad (5.19)$$

which implies, from (5.15), that

$$\mathbb{E}\{F(E_{\tau_{\varpi}}, Y(\tau_{\varpi}), \alpha_{\tau_{\varpi}})1_{\{\tau_{\varpi} \wedge < \infty\} \cap \{\tau_h = \infty\}}\} \leq \mathbb{E}\{F(E_{\tau_{\mu}}, Y(\tau_{\mu}), \alpha_{\tau_{\mu}})1_{\{\tau_{\mu} < \tau_h \wedge \bar{\theta}\}}\}. \quad (5.20)$$

Furthermore, we define

$$A_{\mu} := \sup\{F(t_2, y, \alpha) : (t_2, y, \alpha) \in \mathbb{R}^+ \times \bar{\mathcal{A}}_{\mu} \times \mathcal{M}_2(\mathbb{R}^n)\}, \quad (5.21)$$

with  $\lim_{\mu \rightarrow 0} A_{\mu} = 0$  according to assumption (1). So, there exists a small  $\mu$  such that

$$\frac{A_{\mu}}{\phi(\varpi)} < \frac{\varepsilon}{5}. \quad (5.22)$$

From (5.20)–(5.22),

$$\mathbb{P}\{\{\tau_{\varpi} < \infty\} \cap \{\tau_h = \infty\}\} \leq \frac{A_{\mu}}{\phi(\varpi)} < \frac{\varepsilon}{5}. \quad (5.23)$$

Similar to (5.15), it is easy to derive

$$\mathbb{P}\{\{\tau_{\varpi} < \infty\} \cap \{\tau_h = \infty\}\} \geq \mathbb{P}\{\tau_{\varpi} < \infty\} - \mathbb{P}\{\tau_h < \infty\} \geq \mathbb{P}\{\tau_{\varpi} < \infty\} - \frac{\varepsilon}{5}. \quad (5.24)$$

From (5.23) and (5.24), we obtain

$$\mathbb{P}\{\tau_{\varpi} < \infty\} < \frac{2\varepsilon}{5}. \quad (5.25)$$

Combining (5.15) and (5.25), we obtain

$$\mathbb{P}\{\varrho < \infty \text{ and } \tau_{\varpi} = \infty\} \geq \mathbb{P}\{\varrho < \infty\} - \mathbb{P}\{\tau_{\varpi} < \infty\} > 1 - \varepsilon,$$

which indicates

$$\mathbb{P}\{\omega : \limsup_{t \rightarrow \infty} |Y(t, y_0)| \leq \varpi\} > 1 - \varepsilon.$$

Lastly, since  $\varpi$  is arbitrary, we put  $\varpi \rightarrow 0$  to arrive at

$$\mathbb{P}\{\omega : \lim_{t \rightarrow \infty} |Y(t, y_0)| = 0\} \geq 1 - \varepsilon.$$

The proof is then complete.  $\square$

**Definition 5.4.** ([31, 45]) Assume for all  $y_0 \in \mathbb{R}^n$  the trivial solution of (1.2) is stable in probability such that

$$\mathbb{P}\{\lim_{t \rightarrow \infty} |Y(t, y_0)| = 0\} = 1.$$

Then it is called globally asymptotically stable in probability or stochastically asymptotically stable in large.

**Theorem 5.4.** Suppose for all  $(t_2, y, \alpha) \in \mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n)$ , there exists a function  $F(t_2, y, \alpha) \in \mathcal{P}_{b,+}(\mathbb{R}^+ \times \mathbb{R}^n \times \mathcal{M}_2(\mathbb{R}^n))$  with  $h > 2c$  and  $\phi \in \mathcal{H}$  such that

- (1)  $F(t_2, 0, \delta_0) = 0$ ,
- (2)  $\phi(|y|) \leq F(t_2, y, \alpha)$ ,
- (3)  $\lim_{|y| \rightarrow \infty} \inf_{t_2 \geq 0, \alpha \in \mathcal{M}_2(\mathbb{R}^n)} F(t_2, y, \alpha) = \infty$ ,
- (4)  $\mathbb{L}_{f,g,\tilde{h}} F(t_2, y, \alpha) \leq -\gamma(y)$  a.s.,

where  $\gamma(y) \geq 0$  and  $\gamma(y) \neq 0$  when  $y \neq 0$ . Then the trivial solution of (1.2) is globally asymptotically stable in probability.

*Proof.* The proof uses a similar idea as Theorem 4 in [40], so we omit it here.  $\square$

We close this section by giving an example to enhance the theoretical results.

**Example 5.1.** Consider the following one-dimensional mean-field stochastic differential equation driven by time-changed Lévy noise:

$$\begin{aligned} dY(t) = & -b(t, E_t) \sin\left(\int_{\mathbb{R}} x\alpha_t(dx)\right) dE_t + \sigma(t, E_t) \sqrt{\left|\int_{\mathbb{R}} x\alpha_t(dx)\right|} dW_{E_t} \\ & + \int_{|z|<c} \sigma_1(t, E_t, z) Y(t) \tilde{N}(dE_t, dz), \end{aligned} \quad (5.26)$$

where

$$\begin{aligned} f(t_1, t_2, y, \alpha) &= -b(t_1, t_2) \sin\left(\int_{\mathbb{R}} x\alpha(dx)\right), \\ g(t_1, t_2, y, \alpha) &= \sigma(t_1, t_2) \sqrt{\left|\int_{\mathbb{R}} x\alpha(dx)\right|}, \\ \tilde{h}(t_1, t_2, y, \alpha, z) &= \sigma_1(t_1, t_2, z)y, \end{aligned}$$

with initial value  $Y(0) = y_0$ , where  $b(t_1, t_2)$ ,  $\sigma(t_1, t_2)$ , and  $\sigma_1(t_1, t_2, z)$  are real-valued and  $\mathcal{G}_t$ -measurable functions satisfying

$$0 < b^2(t_1, t_2), \sigma^2(t_1, t_2), \sigma_1^2(t_1, t_2, z) \leq K,$$

for all  $t_1, t_2 \in \mathbb{R}^+ \times \mathbb{R}^+$  and  $K > 0$ . So, one can justify that for  $(t_1, t_2, y, \alpha) \in \mathbb{R}^+ \times \mathbb{R}^+ \times \mathbb{R} \times \mathcal{M}_2(\mathbb{R})$  and using the fact that  $\sin(l) \leq l$ , we have

$$|f(t_1, t_2, y, \alpha)|^2 + \|g(t_1, t_2, y, \alpha)\|^2 + \int_{|z|<c} |\tilde{h}(t_1, t_2, y, \alpha, z)|^2 \nu(dz)$$

$$\begin{aligned}
&\leq K\left\{\left|\sin\left(\int_{\mathbb{R}}x\alpha(dx)\right)\right|^2 + \left|\int_{\mathbb{R}}x\alpha(dx)\right| + |y|^2\right\} \\
&\leq K\left\{\left|\sin\left(\int_{\mathbb{R}}x\alpha(dx)\right)\right| + \left|\int_{\mathbb{R}}x\alpha(dx)\right| + |y|^2\right\} \\
&\leq K\left\{|y|^2 + 2\left|\int_{\mathbb{R}}x\alpha(dx)\right|\right\} \\
&\leq K\left\{1 + |y|^2 + \left|\int_{\mathbb{R}}x\alpha(dx)\right|^2\right\} \\
&\leq K\{1 + |y|^2 + \rho_2^2(\alpha, \delta_0)\}.
\end{aligned}$$

Moreover, for  $(y_1, \alpha_1), (y_2, \alpha_2) \in \mathbb{R} \times \mathcal{M}_2(\mathbb{R})$ ,

$$\begin{aligned}
&|f(t_1, t_2, y_1, \alpha_1) - f(t_1, t_2, y_2, \alpha_2)|^2 + \|g(t_1, t_2, y_1, \alpha_1) - g(t_1, t_2, y_2, \alpha_2)\|^2 \\
&\quad + \int_{|z|<c} |\tilde{h}(t_1, t_2, y_1, \alpha_1, z) - \tilde{h}(t_1, t_2, y_2, \alpha_2, z)|^2 \nu(dz) \\
&\leq K\left\{|y_1 - y_2|^2 + \left|\sqrt{\left|\int_{\mathbb{R}}x\alpha_1(dx)\right|} - \sqrt{\left|\int_{\mathbb{R}}x\alpha_2(dx)\right|}\right|^2\right. \\
&\quad \left. + \left|\sin\left(\int_{\mathbb{R}}x\alpha_1(dx)\right) - \sin\left(\int_{\mathbb{R}}x\alpha_2(dx)\right)\right|^2\right\} \\
&\leq K\left\{|y_1 - y_2|^2 + \left|\sqrt{\left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|}\right|^2\right. \\
&\quad \left. + 4\left|\sin\frac{\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)}{2}\right|^2\right\} \\
&\leq K\left\{|y_1 - y_2|^2 + \left|\sqrt{\left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|}\right|^2\right. \\
&\quad \left. + \left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|^2\right\} \\
&\leq K\left\{|y_1 - y_2|^2 + \left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|^2\right. \\
&\quad \left. + \kappa^2\left(\left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|\right)\right\},
\end{aligned}$$

where we have used the fact that  $\kappa(u) = \sqrt{u} \in [0, \infty)$ , which is increasing with  $\kappa(0) = 0$  and satisfying  $\int_{0^+} \frac{1}{\kappa(u)} du = \infty$ . Now, taking  $\hat{\kappa}(u) = u + \kappa^2(u^{\frac{1}{2}})$ , we obtain

$$\begin{aligned}
&|f(t_1, t_2, y_1, \alpha_1) - f(t_1, t_2, y_2, \alpha_2)|^2 + \|g(t_1, t_2, y_1, \alpha_1) - g(t_1, t_2, y_2, \alpha_2)\|^2 \\
&\quad + \int_{|z|<c} |\tilde{h}(t_1, t_2, y_1, \alpha_1, z) - \tilde{h}(t_1, t_2, y_2, \alpha_2, z)|^2 \nu(dz) \\
&\leq K\left\{|y_1 - y_2|^2 + \left|\int_{\mathbb{R}}x\alpha_1(dx) - \int_{\mathbb{R}}x\alpha_2(dx)\right|^2\right\}
\end{aligned}$$

$$\begin{aligned}
& + \kappa^2 \left( \left| \int_{\mathbb{R}} x\alpha_1(dx) - \int_{\mathbb{R}} x\alpha_2(dx) \right|^2 \right)^{\frac{1}{2}} \Big\} \\
& \leq K \left\{ |y_1 - y_2|^2 + \hat{\kappa} \left( \left| \int_{\mathbb{R}} x\alpha_1(dx) - \int_{\mathbb{R}} x\alpha_2(dx) \right|^2 \right) \right\} \\
& \leq K \{ |y_1 - y_2|^2 + \hat{\kappa}(\rho_2^2(\alpha_1, \alpha_2)) \},
\end{aligned}$$

where  $\hat{\kappa}(u)$  is a non-decreasing function on  $\mathbb{R}^+$  and  $\hat{\kappa}(0) = 0$ . Moreover, we can obtain

$$\int_{0^+} \frac{ds}{\hat{\kappa}(s)} = \int_{0^+} \frac{1}{\kappa^2(s^{\frac{1}{2}}) + s} ds = \int_{0^+} \frac{u}{\kappa^2(u) + u^2} du = \infty.$$

Thus, the coefficients  $f, g, \hbar$  satisfy Conditions 3.1–3.3, which yields that Eq (5.26) has a strong solution.

For some  $0 < \beta < 1$ , we define

$$F(t_2, y, \alpha) = \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha(dx) \right) \right) |y|^\beta,$$

and it is easily seen that

$$\begin{aligned}
\partial_{t_2} F(t_2, y, \alpha) &= 0, & F_y(t_2, y, \alpha) &= \beta \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha(dx) \right) \right) |y|^{\beta-1}, \\
F_{yy}(t_2, y, \alpha) &= \beta(\beta - 1) \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha(dx) \right) \right) |y|^{\beta-2}, \\
\partial_\alpha F(t_2, y, \alpha)(v) &= -\sin \left( \int_{\mathbb{R}} x\alpha(dx) \right) |y|^\beta, & \partial_v \partial_\alpha F(t_2, y, \alpha)(v) &= 0.
\end{aligned}$$

Hence, it holds that

$$\begin{aligned}
& \mathbb{L}_{f,g,\hbar} F(t_2, y, \alpha) \\
&= -\beta \sin \left( \int_{\mathbb{R}} x\alpha(dx) \right) \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha_0(dx) \right) \right) b(t_1, t_2) |y|^{\beta-1} \\
&+ \frac{\beta(\beta - 1)}{2} \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha_0(dx) \right) \right) \sigma^2(t_1, t_2) \left| \int_{\mathbb{R}} x\alpha(dx) \right| |y|^{\beta-2} \\
&+ \sin^2 \left( \int_{\mathbb{R}} x\alpha(dx) \right) b(t_1, t_2) \left( \int_{\mathbb{R}} x\alpha(dx) \right) |y|^\beta \\
&+ \int_{|z|<c} \int_0^1 \left[ \left( -\sin \left( \int_{\mathbb{R}} x\alpha(dx) \right) \right) |y|^\beta (y + \xi \sigma_1(t_1, t_2, z)y) \right. \\
&\quad \left. + \sin \left( \int_{\mathbb{R}} x\alpha(dx) \right) |y|^\beta \left( \int_{\mathbb{R}} x\alpha(dx) \right) \sigma_1(t_1, t_2, z)y \right] d\xi v(dz) \\
&+ \int_{|z|<c} \left[ \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha_0(dx) \right) \right) |y + \sigma_1(t_1, t_2, z)y|^\beta \right. \\
&\quad \left. - \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha_0(dx) \right) \right) |y|^\beta - \beta \left( 1 + \cos \left( \int_{\mathbb{R}} x\alpha_0(dx) \right) \right) \right. \\
&\quad \left. \times \sigma_1(t_1, t_2, z) |y|^\beta \right] v(dz).
\end{aligned}$$

Therefore, if

$$\left\{ \begin{array}{l}
 b(t_1, t_2) \left( \int_{\mathbb{R}} x \alpha(dx) \right) \sin^2 \left( \int_{\mathbb{R}} x \alpha(dx) \right) + \left( 1 + \cos \left( \int_{\mathbb{R}} x \alpha_0(dx) \right) \right) \\
 \quad \times \int_{|z| < c} [|1 + \sigma_1(t_1, t_2, z)|^\beta - 1 - \beta \sigma_1(t_1, t_2, z)] \nu(dz) \leq 0, \\
 \int_{|z| < c} \sin \left( \int_{\mathbb{R}} x \alpha(dx) \right) \left( \int_{\mathbb{R}} x \alpha(dx) \right) \sigma_1(t_1, t_2, z) \nu(dz) \leq 0, \\
 \int_{|z| < c} \int_0^1 \sin \left( \int_{\mathbb{R}} x \alpha(dx) \right) (1 + \xi \sigma_1(t_1, t_2, z)) \sigma_1(t_1, t_2, z) d\xi \nu(dz) \geq 0, \\
 \beta b(t_1, t_2) \sin \left( \int_{\mathbb{R}} x \alpha(dx) \right) \left( 1 + \cos \left( \int_{\mathbb{R}} x \alpha_0(dx) \right) \right) \geq 0, \\
 \left. \frac{\beta(\beta-1)}{2} \sigma^2(t_1, t_2) \left| \int_{\mathbb{R}} x \alpha(dx) \right| \left( 1 + \cos \left( \int_{\mathbb{R}} x \alpha_0(dx) \right) \right) \leq 0 \right.
 \end{array} \right. \quad (5.27)$$

hold a.s. for all  $(t_1, t_2) \in (J, \mathbb{R}^+)$ , then from Theorem 5.2, it is known that the trivial solution of Eq (5.26) is stable in probability. If inequalities (5.27) hold strictly, Theorem 5.4 guarantees that the trivial solution of Eq (5.26) is globally asymptotically stable in probability.

## 6. Conclusions

In this work, we have investigated a new class of MFSDs perturbed by time-changed Lévy noise. To show the existence and uniqueness of solutions, we have derived a new time-changed retarded integral inequality and used the Carathéodory approximations and non-Lipschitz condition, generalizing some existing results. Second, the classical Itô formula is extended from SDEs to MFSDs perturbed by time-changed Lévy noise. Then, to show the applicability of Itô formula, we have studied the  $p$ -th moment asymptotic stability, stability in probability, asymptotic stability in probability, and globally asymptotic stability in probability for the trivial solution depending on the Lyapunov function. Finally, the theoretical results are validated through an example. For the future investigations, we may study the practical stability with respect to part of the variables of MFSDs perturbed by time-changed Lévy noise.

### Author contributions

M. Abouagwa: Validation, writing—original draft, supervision, conceptualization, writing—review and editing; M. I. Tawdrous: Validation, writing—review and editing. All authors have read and agreed to the revised version of the manuscript.

### Use of Generative-AI tools declaration

The authors declare they have not used Artificial Intelligence (AI) tools in the creation of this article.

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## Conflict of interest

The authors declare no conflicts.

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