

Research Article

Stationary in Distributions of Numerical Solutions for Stochastic Partial Differential Equations with Markovian Switching

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We investigate a class of stochastic partial differential equations with Markovian switching. By using the Euler-Maruyama scheme both in time and in space of mild solutions, we derive sufficient conditions for the existence and uniqueness of the stationary distributions of numerical solutions. Finally, one example is given to illustrate the theory.

1. Introduction

The theory of numerical solutions of stochastic partial differential equations (SPDEs) has been well developed by many authors [1–5]. In [2], Debussche considered the error of the Euler scheme for the nonlinear stochastic partial differential equations by using Malliavin calculus. Gyöngy and Millet [3] discussed the convergence rate of space time approximations for stochastic evolution equations. Shardlow [5] investigated the numerical methods of the mild solutions for stochastic parabolic PDEs derived by space-time white noise by applying finite difference approach.

On the other hand, the parameters of SPDEs may experience abrupt changes caused by phenomena such as component failures or repairs, changing subsystem interconnections, and abrupt environmental disturbances [6–9], and the continuous-time Markov chains have been used to model these parameter jumps. An important equation is a class of SPDEs with Markovian switching

$$\begin{aligned} dX(t) = & [AX(t) + f(X(t), r(t))] dt \\ & + g(X(t), r(t)) dW(t), \quad t \geq 0. \end{aligned} \quad (1)$$

Here the state vector has two components $X(t)$ and $r(t)$, the first one is normally referred to as the state while the second one is regarded as the mode. In its operation, the system will switch from one mode to another one in a random way, and

the switching among the modes is governed by the Markov chain $r(t)$.

Since only a few SPDEs with Markovian switching have explicit formulae, numerical (approximate) schemes of SPDEs with Markovian switching are becoming more and more popular. In this paper, we will study the stationary distribution of numerical solutions of SPDEs with Markovian switching. Bao et al. [10] investigated the stability in distribution of mild solutions to SPDEs. Bao and Yuan [11] discussed the numerical approximation of stationary distribution for SPDEs. For the stationary distribution of numerical solutions of stochastic differential equations in finite-dimensional space, Mao et al. [12] utilized the Euler-Maruyama scheme with variable step size to obtain the stationary distribution and they also proved that the probability measures induced by the numerical solutions converge weakly to the stationary distribution of the true solution. But since the mild solutions of SPDEs with Markovian switching do not have stochastic differential, a significant consequence of this fact is that the Itô formula cannot be used for mild solutions of SPDEs with Markovian switching directly. Consequently, we generalize the stationary distribution of numerical solutions of the finite dimensional stochastic differential equations with Markovian switching to that of infinite dimensional cases.

Motivated by [11–13], we will show in this paper that the mild solutions of SPDE with Markovian switching (1) have a unique stationary distribution for sufficiently small step size.

So this paper is organised as follows: in Section 2, we give necessary notations and define Euler-Maruyama scheme of mild solutions. In Section 3, we give some lemmas and the main result in this paper. Finally, we will give an example to illustrate the theory in Section 4.

2. Statements of Problem

Throughout this paper, unless otherwise specified, we let $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$ be complete probability space with a filtration $\{\mathcal{F}_t\}_{t \geq 0}$ satisfying the usual conditions (i.e., it is increasing and right continuous while \mathcal{F}_0 contains all \mathbb{P} -null sets). Let $(H, \langle \cdot, \cdot \rangle_H, \| \cdot \|_H)$ be a real separable Hilbert space and $W(t)$ an H -valued cylindrical Brownian motion (Wiener process) defined on the probability space. Let I_G be the indicator function of a set G . Denote by $(\mathcal{L}(H), \| \cdot \|)$ and $(\mathcal{L}_{HS}(H), \| \cdot \|_{HS})$ the family of bounded linear operators and Hilbert-Schmidt operator from H into H , respectively. Let $r(t), t \geq 0$, be a right-continuous Markov chain on the probability space taking values in a finite state space $\mathbb{S} = \{1, 2, \dots, N\}$ with the generator $\Gamma = (\gamma_{ij})_{N \times N}$ given by

$$\mathbb{P} \{r(t + \delta) = j \mid r(t) = i\} = \begin{cases} \gamma_{ij}\delta + o(\delta) & \text{if } i \neq j, \\ 1 + \gamma_{ij}\delta + o(\delta) & \text{if } i = j, \end{cases} \quad (2)$$

where $\delta > 0$. Here $\gamma_{ij} > 0$ is the transition rate from i to j if $i \neq j$ while

$$\gamma_{ii} = -\sum_{j \neq i} \gamma_{ij}. \quad (3)$$

We assume that the Markov chain $r(\cdot)$ is independent of the Brownian motion $W(\cdot)$. It is well known that almost every sample path of $r(\cdot)$ is a right-continuous step function with finite number of simple jumps in any finite subinterval of $\mathbb{R}_+ := [0, +\infty)$.

Consider SPDEs with Markovian switching on H

$$dX(t) = [AX(t) + f(X(t), r(t))] dt + g(X(t), r(t)) dW(t), \quad t \geq 0, \quad (4)$$

with initial value $X(0) = x \in H$ and $r(0) = i \in \mathbb{S}$. Here $f : H \times \mathbb{S} \rightarrow H, g : H \times \mathbb{S} \rightarrow \mathcal{L}_{HS}(H)$. Throughout the paper, we impose the following assumptions.

(A1) $(A, \mathcal{D}(A))$ is a self-adjoint operator on H generating a C_0 -semigroup $\{e^{At}\}_{t \geq 0}$, such that $\|e^{At}\| \leq e^{-\alpha t}$ for some $\alpha > 0$. In this case, $-A$ has discrete spectrum $0 < \rho_1 \leq \rho_2 \leq \dots \leq \lim_{i \rightarrow \infty} \rho_i = \infty$ with corresponding eigenbasis $\{e_i\}_{i \geq 1}$ of H .

(A2) Both f and g are globally Lipschitz continuous. That is, there exists a constant $L > 0$ such that

$$\|f(x, j) - f(y, j)\|_H^2 \vee \|g(x, j) - g(y, j)\|_{HS}^2 \leq L\|x - y\|_H^2, \quad \forall x, y \in H, j \in \mathbb{S}; \quad (5)$$

(A3) There exist $\mu > 0$ and $\lambda_j > 0, (j = 1, 2, \dots, N)$ such that

$$2\lambda_j \langle x - y, f(x, j) - f(y, j) \rangle_H + \lambda_j \|g(x, j) - g(y, j)\|_{HS}^2 + \sum_{l=1}^N \gamma_{jl} \lambda_l \|x - y\|_H^2 \leq -\mu \|x - y\|_H^2, \quad \forall x, y \in H, j \in \mathbb{S}. \quad (6)$$

It is well known (see [1, 8]) that under (A1)–(A3), (4) has a unique mild solution $X(t)$ on $t \geq 0$. That is, for any $X(0) = x \in H$ and $r(0) = i \in \mathbb{S}$, there exists a unique H -valued adapted process $X(t)$ such that

$$X(t) = e^{tA}x + \int_0^t e^{(t-s)A} f(X(s), r(s)) ds + \int_0^t e^{(t-s)A} g(X(s), r(s)) dW(s). \quad (7)$$

Moreover, the pair $Z(t) = (X(t), r(t))$ is a time-homogeneous Markov process.

Remark 1. We observe that (A2) implies the following linear growth conditions:

$$\|f(x, j)\|_H^2 \vee \|g(x, j)\|_{HS}^2 \leq \bar{L}(1 + \|x\|_H^2), \quad \forall x \in H, j \in \mathbb{S}, \quad (8)$$

where $\bar{L} = 2 \max_{j \in \mathbb{S}} (L \vee \|f(0, j)\|_H^2 \vee \|g(0, j)\|_{HS}^2)$.

Remark 2. We also establish another property from (A3):

$$\begin{aligned} 2\lambda_j \langle x, f(x, j) \rangle_H + \lambda_j \|g(x, j)\|_{HS}^2 + \sum_{l=1}^N \gamma_{jl} \lambda_l \|x\|_H^2 &\leq 2\lambda_j \langle x, f(x, j) - f(0, j) \rangle_H \\ &+ \lambda_j \|g(x, j) - g(0, j)\|_{HS}^2 + \sum_{l=1}^N \gamma_{jl} \lambda_l \|x\|_H^2 \\ &+ 2\lambda_j \langle x, f(0, j) \rangle_H \\ &+ 2\lambda_j \langle g(x, j) - g(0, j), g(0, j) \rangle_{HS} + \lambda_j \|g(0, j)\|_{HS}^2 \\ &\leq -\mu \|x\|_H^2 + \frac{\mu}{4} \|x\|_H^2 + \frac{4\lambda_j^2 \|f(0, j)\|_H}{\mu} \\ &+ \frac{\mu}{4L} \|g(x, j) - g(0, j)\|_{HS}^2 \\ &+ \frac{4L\lambda_j^2}{\mu} \|g(0, j)\|_{HS}^2 + \lambda_j \|g(0, j)\|_{HS}^2 \\ &\leq -\mu \|x\|_H^2 + \frac{\mu}{4} \|x\|_H^2 + \frac{\mu}{4} \|x\|_H^2 + \frac{4\lambda_j^2 \|f(0, j)\|_H}{\mu} \end{aligned}$$

$$\begin{aligned}
 & + \frac{4L\lambda_j^2}{\mu} \|g(0, j)\|_{HS}^2 + \lambda_j \|g(0, j)\|_{HS}^2 \\
 & \leq -\frac{\mu}{2} \|x\|_H^2 + \alpha_1, \quad \forall x \in H, j \in \mathbb{S},
 \end{aligned} \tag{9}$$

where $\alpha_1 := \max_{j \in \mathbb{S}} [(4\lambda_j^2 \|f(0, j)\|_H^2 / \mu) + (4L\lambda_j^2 / \mu) \|g(0, j)\|_{HS}^2 + \lambda_j \|g(0, j)\|_{HS}^2]$ and $\langle T, S \rangle_{HS} := \sum_{i=1}^{\infty} \langle Te_i, Se_j \rangle_H$ for $S, T \in \mathcal{L}_{HS}(H)$.

Denote by $Z^{x,i}(t) = (X^{x,i}(t), r^i(t))$ the mild solution of (4) starting from $(x, i) \in H \times \mathbb{S}$. For any subset $A \in \mathfrak{B}(H), B \subset \mathbb{S}$, let $\mathbb{P}_t((x, i), A \times B)$ be the probability measure induced by $Z^{x,i}(t), t \geq 0$. Namely,

$$\mathbb{P}_t((x, i), A \times B) = \mathbb{P}(Z^{x,i} \in A \times B), \tag{10}$$

where $\mathfrak{B}(H)$ is the family of the Borel subset of H .

Denote by $\mathcal{P}(H \times \mathbb{S})$ the family by all probability measures on $H \times \mathbb{S}$. For $P_1, P_2 \in \mathcal{P}(H \times \mathbb{S})$, define the metric $d_{\mathbb{L}}$ as follows:

$$\begin{aligned}
 & d_{\mathbb{L}}(P_1, P_2) \\
 & = \sup_{\varphi \in \mathbb{L}} \left| \sum_{j=1}^N \int_H \varphi(u, j) P_1(du, j) - \sum_{j=1}^N \int \varphi(u, j) P_2(du, j) \right|,
 \end{aligned} \tag{11}$$

where $\mathbb{L} = \{\varphi : H \times \mathbb{S} \rightarrow \mathbb{R} : |\varphi(u, j) - \varphi(v, l)| \leq \|u - v\|_H + |j - l|, \text{ and } |\varphi(u, j)| \leq 1, \text{ for } u, v \in K, j, l \in \mathbb{S}\}$.

Remark 3. It is known that the weak convergence of probability measures is a metric concept with respect to classes of test function. In other words, a sequence of probability measures $\{P_k\}_{k \geq 1}$ of $\mathcal{P}(H \times \mathbb{S})$ converges weakly to a probability measure $P_0 \in \mathcal{P}(H \times \mathbb{S})$ if and only if $\lim_{k \rightarrow \infty} d_{\mathbb{L}}(P_k, P_0) = 0$.

Definition 4. The mild solution $Z(t) = (X(t), r(t))$ of (4) is said to have a stationary distribution $\pi(\cdot \times \cdot) \in \mathcal{P}(H \times \mathbb{S})$ if the probability measure $\mathbb{P}_t((x, i), (\cdot \times \cdot))$ converges weakly to $\pi(\cdot \times \cdot)$ as $t \rightarrow \infty$ for every $i \in \mathbb{S}$, and every $x \in U$, a bounded subset of H , that is,

$$\begin{aligned}
 & \lim_{t \rightarrow \infty} d_{\mathbb{L}}(\mathbb{P}_t(x, i), \pi(\cdot \times \cdot)) \\
 & = \lim_{t \rightarrow \infty} \left(\sup_{\varphi \in \mathbb{L}} \left| \mathbb{E} \varphi(Z^{x,i}(t)) - \sum_{j=1}^N \int_H \varphi(u, j) \pi(du, j) \right| \right) = 0.
 \end{aligned} \tag{12}$$

By Theorem 3.1 in [10] and Theorem 3.1 in [14], we have the following.

Theorem 5. Under (A1)–(A3), the Markov process $Z(t)$ has a unique stationary distribution $\pi(\cdot \times \cdot) \in \mathcal{P}(H \times \mathbb{S})$.

For any $n \geq 1$, let $\pi_n : H \rightarrow H_n := \text{Span}\{e_1, e_2, \dots, e_n\}$ be the orthogonal projection. Consider SPDEs with Markovian switching on H_n ,

$$\begin{aligned}
 dX^n(t) & = [A_n X^n(t) + f_n(X^n(t), r(t))] dt \\
 & + g_n(X^n(t), r(t)) dW(t),
 \end{aligned} \tag{13}$$

with initial data $X^n(0) = \pi_n x = \sum_{i=1}^n \langle x, e_i \rangle_H e_i, x \in H$. Here $A_n = \pi_n A, f_n = \pi_n f, g_n = \pi_n g$.

Therefore, we can observe that

$$\begin{aligned}
 A_n x & = Ax, \quad e^{tA_n x} = e^{tAx}, \quad \langle x, f_n \rangle_H = \langle x, f \rangle_H, \\
 \langle x, g_n \rangle_H & = \langle x, g \rangle_H, \quad \forall x \in H_n,
 \end{aligned} \tag{14}$$

By the property of the projection operator and (A2), we have

$$\begin{aligned}
 & \|A_n(x - y)\|_H^2 \vee \|f_n(x, j) - f_n(y, j)\|_H^2 \\
 & \vee \|g_n(x, j) - g_n(y, j)\|_{HS}^2 \\
 & \leq \lambda_n^2 \|x - y\|_H^2 \vee \|f(x, j) - f(y, j)\|_H^2 \\
 & \vee \|g(x, j) - g(y, j)\|_{HS}^2 \leq (\lambda_n^2 \vee L) \|x - y\|_H^2, \\
 & \quad \forall x, y \in H_n, j \in \mathbb{S}.
 \end{aligned} \tag{15}$$

Hence, (13) admits a unique strong solution $\{X^n(t)\}_{t \geq 0}$ on H_n (see [8]).

We now introduce an Euler-Maruyama based computational method. The method makes use of the following lemma (see [15]).

Lemma 6. Given $\Delta > 0$, then $\{r(k\Delta), k = 0, 1, 2, \dots\}$ is a discrete Markov chain with the one-step transition probability matrix

$$P(\Delta) = (P_{i,j}(\Delta))_{N \times N} = e^{\Delta \Gamma}. \tag{16}$$

Given a fixed step size $\Delta > 0$ and the one-step transition probability matrix $P(\Delta)$ in (16), the discrete Markov chain $\{r(k\Delta), k = 0, 1, 2, \dots\}$ can be simulated as follows: let $r(0) = i_0$, and compute a pseudorandom number ξ_1 from the uniform (0, 1) distribution.

Define

$$r(\Delta) = \begin{cases} i, & i \in \mathbb{S} - \{N\} \\ & \text{such that } \sum_{j=1}^{i-1} P_{r(0),j}(\Delta) \leq \xi_1 \\ & < \sum_{j=1}^i P_{r(0),j}(\Delta), \\ N, & \sum_{j=1}^{N-1} P_{r(0),j}(\Delta) \leq \xi_1, \end{cases} \tag{17}$$

where we set $\sum_{j=1}^0 P_{r(0),j}(\Delta) = 0$ as usual. Having computed $r(0), r(\Delta), \dots, r(k\Delta)$, we can compute $r((k+1)\Delta)$ by drawing a uniform $(0, 1)$ pseudorandom number ξ_{k+1} and setting

$$r((k+1)\Delta) = \begin{cases} i, & i \in \mathbb{S} - \{N\} \\ & \text{such that } \sum_{j=1}^{i-1} P_{r(k\Delta),j}(\Delta) \\ & \leq \xi_{k+1} < \sum_{j=1}^i P_{r(k\Delta),j}(\Delta), \\ N, & \sum_{j=1}^{N-1} P_{r(k\Delta),j}(\Delta) \leq \xi_{k+1}. \end{cases} \quad (18)$$

The procedure can be carried out repeatedly to obtain more trajectories.

We now define the Euler-Maruyama approximation for (13). For a stepsize $\Delta \in (0, 1)$, the discrete approximation $\bar{Y}^n(k\Delta) \approx X^n(k\Delta)$, is formed by simulating from $\bar{Y}^n(0) = \pi_n x, r(0) = r_0$, and

$$\begin{aligned} \bar{Y}^n((k+1)\Delta) &= e^{\Delta A_n} \{ \bar{Y}^n(k\Delta) + f_n(\bar{Y}^n(k\Delta), r(k\Delta)) \Delta \\ &\quad + g_n(\bar{Y}^n(k\Delta), r(k\Delta)) \Delta W_k \}, \end{aligned} \quad (19)$$

where $\Delta W_k = W((k+1)\Delta) - W(k\Delta)$.

To carry out our analysis conveniently, we give the continuous Euler-Maruyama approximation solution which is defined by

$$\begin{aligned} Y^n(t) &= e^{tA_n} \pi_n x + \int_0^t e^{(t-[s])A_n} f_n(Y^n([s]), r([s])) ds \\ &\quad + \int_0^t e^{(t-[s])A_n} g_n(Y^n([s]), r([s])) dW(s) \\ &= e^{tA_n} \pi_n x + \int_0^t e^{(t-[s])A} f_n(Y^n([s]), r([s])) ds \\ &\quad + \int_0^t e^{(t-[s])A} g_n(Y^n([s]), r([s])) dW(s), \end{aligned} \quad (20)$$

where $[t] = [t/\Delta]\Delta$ and $[t/\Delta]$ denotes the integer part of t/Δ and $Y^n(0) = \bar{Y}^n(0) = \pi_n x$, and $Y^n(k\Delta) = \bar{Y}^n(k\Delta)$.

It is obvious that $Y^n(t)$ coincides with the discrete approximation solution at the gridpoints. For any Borel set $A \in \mathfrak{B}(H_n)$, $x \in H_n, i, j \in \mathbb{S}$, let $\bar{Z}^n(k\Delta) = (\bar{Y}^n(k\Delta), r(k\Delta))$,

$$\begin{aligned} \mathbb{P}^{n,\Delta}((x, i), A \times \{j\}) &:= \mathbb{P}(\bar{Z}^n(\Delta) \in A \times \{j\} \mid \bar{Z}^n(0) = (x, i)), \\ \mathbb{P}_k^{n,\Delta}((x, i), A \times \{j\}) &:= \mathbb{P}(\bar{Z}^n(k\Delta) \in A \times \{j\} \mid \bar{Z}^n(0) = (x, i)). \end{aligned} \quad (21)$$

Following the argument of Theorem 5 in [13], we have the following.

Lemma 7. $\{\bar{Z}^n(k\Delta)\}_{k \geq 0}$ is a homogeneous Markov process with the transition probability kernel $\mathbb{P}^{n,\Delta}((x, i), A \times \{j\})$.

To highlight the initial value, we will use notation $\{\bar{Z}^{n,(x,i)}(k\Delta)\}$.

Definition 8. For a given stepsize $\Delta > 0$, $\{\bar{Z}^{n,(x,i)}(k\Delta)\}_{k \geq 0}$ is said to have a stationary distribution $\{\pi^{n,\Delta}(\cdot \times \cdot)\} \in \mathcal{P}(H_n \times \mathbb{S})$ if the k -step transition probability kernel $\mathbb{P}_k^{n,\Delta}((x, i), \cdot \times \cdot)$ converges weakly to $\pi^{n,\Delta}(\cdot \times \cdot)$ as $k \rightarrow \infty$, for every $(x, i) \in H_n \times \mathbb{S}$, that is,

$$\lim_{k \rightarrow \infty} d_{\perp} (P_k^{n,\Delta}((x, i), \cdot \times \cdot), \pi^{n,\Delta}(\cdot \times \cdot)) = 0. \quad (22)$$

We will establish our result of this paper in Section 3.

Theorem 9. Under (A1)–(A3), for a given stepsize $\Delta > 0$, and arbitrary $x \in H_n, i \in \mathbb{S}$, $\{\bar{Z}^{n,(x,i)}(k\Delta)\}_{k \geq 0}$ has a unique stationary distribution $\pi^{n,\Delta}(\cdot \times \cdot) \in \mathcal{P}(H_n \times \mathbb{S})$.

3. Stationary in Distribution of Numerical Solutions

In this section, we shall present some useful lemmas and prove Theorem 9. In what follows, $C > 0$ is a generic constant whose values may change from line to line.

For any initial value (x, i) , let $Y^{n,x,i}(t)$ be the continuous Euler-Maruyama solution of (20) and starting from $(x, i) \in H \times \mathbb{S}$. Let $X^{x,i}(t)$ be the mild solution of (4) and starting from $(x, i) \in H \times \mathbb{S}$.

Lemma 10. Under (A1)–(A3), then

$$\begin{aligned} \mathbb{E} \| Y^{n,x,i}(t) - Y^{n,x,i}([t]) \|_H^2 &\leq 3(\rho_n^2 + 2\bar{L}) \Delta (1 + \mathbb{E} \| Y^n([t]) \|_H^2). \end{aligned} \quad (23)$$

Proof. Write $Y^{n,x,i}(t) = Y^n(t), Y^{n,x,i}([t]) = Y^n([t])$. From (20), we have

$$\begin{aligned} Y^n([t]) &= e^{[t]A} \pi_n x + \int_0^{[t]} e^{([t]-[s])A} f_n(Y^n([s]), r([s])) ds \\ &\quad + \int_0^{[t]} e^{([t]-[s])A} g_n(Y^n([s]), r([s])) dW(s). \end{aligned} \quad (24)$$

Thus,

$$\begin{aligned}
 & Y^n(t) - Y^n([t]) \\
 &= e^{(t-[t])A} \\
 &\quad \times \left(e^{[t]A} \pi_n x \right. \\
 &\quad \left. + \int_0^{[t]} e^{([t]-[s])A} \right. \\
 &\quad \quad \times f_n(Y^n([s]), r([s])) ds \\
 &\quad \left. + \int_0^{[t]} e^{([t]-[s])A} \right. \\
 &\quad \quad \times g_n(Y^n([s]), r([s])) dW(s) \left. \right) \\
 &- Y^n([t]) \\
 &+ \int_{[t]}^t e^{(t-[s])A} f_n(Y^n([s]), r([s])) ds \\
 &+ \int_{[t]}^t e^{(t-[s])A} g_n(Y^n([s]), r([s])) dW(s) \\
 &= (e^{(t-[t])A} - \mathbf{1}) Y^n([t]) \\
 &+ \int_{[t]}^t e^{(t-[s])A} f_n(Y^n([s]), r([s])) ds \\
 &+ \int_{[t]}^t e^{(t-[s])A} g_n(Y^n([s]), r([s])) dW(s).
 \end{aligned} \tag{25}$$

Then, by the Hölder inequality and the Itô isometry, we obtain

$$\begin{aligned}
 & \mathbb{E} \| Y^n(t) - Y^n([t]) \|_H^2 \\
 & \leq 3 \left\{ \mathbb{E} \| (e^{(t-[t])A} - \mathbf{1}) Y^n([t]) \|_H^2 \right. \\
 & \quad + \mathbb{E} \int_{[t]}^t \| f_n(Y^n([s]), r([s])) \|_H^2 ds \\
 & \quad \left. + \mathbb{E} \int_{[t]}^t \| g_n(Y^n([s]), r([s])) \|_{HS}^2 ds \right\}.
 \end{aligned} \tag{26}$$

From (A1), we have

$$\begin{aligned}
 & \mathbb{E} \| (e^{(t-[t])A} - \mathbf{1}) Y^n([t]) \|_H^2 \\
 &= \mathbb{E} \left\| \sum_{i=1}^n (e^{-\rho_i(t-[t])} - 1) \langle Y^n([t]), e_i \rangle_H e_i \right\|_H^2 \\
 &\leq (1 - e^{-\rho_n(t-[t])})^2 \mathbb{E} \| Y^n([t]) \|_H^2 \\
 &\leq \rho_n^2 \Delta^2 \mathbb{E} \| Y^n([t]) \|_H^2,
 \end{aligned} \tag{27}$$

here we use the fundamental inequality $1 - e^{-a} \leq a$, $a > 0$. And, by (8), it follows that

$$\begin{aligned}
 & \mathbb{E} \int_{[t]}^t \| f_n(Y^n([s]), r([s])) \|_H^2 ds \\
 & \quad + \mathbb{E} \int_{[t]}^t \| g_n(Y^n([s]), r([s])) \|_{HS}^2 ds \\
 & \leq 2\bar{L}\Delta (1 + \mathbb{E} \| Y^n([t]) \|_H^2).
 \end{aligned} \tag{28}$$

Substituting (27) and (28) into (26), the desired assertion (23) follows. \square

Lemma 11. Under (A1)–(A3), if $\Delta < \min\{1, 1/3(\rho_n^2 + 2\bar{L}), ((4\alpha p + \mu)/(8q + 4q\rho_n^2\bar{L} + 4q\bar{L} + 24q\hat{r} + 6qL(\rho_n^2 + 2\bar{L}))^2)\}$, then there is a constant $C > 0$ that depends on the initial value x but is independent of Δ , such that the continuous Euler-Maruyama solution of (20) has

$$\sup_{t \geq 0} \mathbb{E} \| Y^{n,x,i}(t) \|_H \leq C, \tag{29}$$

where $q = \max_{1 \leq i \leq N} \lambda_i$, $p = \min_{1 \leq i \leq N} \lambda_i$.

Proof. Write $Y^{n,x,i}(t) = Y^n(t)$, $r^i(k\Delta) = r(k\Delta)$. From (20), we have the following differential form:

$$\begin{aligned}
 & dY^n(t) \\
 &= \{AY^n(t) + e^{(t-[t])A} f_n(Y^n([t]), r([t]))\} dt \\
 & \quad + e^{(t-[t])A} g_n(Y^n([t]), r([t])) dW(t),
 \end{aligned} \tag{30}$$

with $Y^n(0) = \pi_n x$.

Let $V(x, i) = \lambda_i \| x \|_H^2$. By the generalised Itô formula, for any $\theta > 0$, we derive from (30) that

$$\begin{aligned}
 & e^{\theta t} \mathbb{E} (\lambda_{r(t)} \| Y^n(t) \|_H^2) \\
 & \leq \lambda_i \| x \|_H^2 + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 & \quad \times \left\{ \theta \| Y^n(s) \|_H^2 + 2 \langle Y^n(s), AY^n(s) \rangle_H \right. \\
 & \quad \left. + 2 \langle Y^n(s), e^{(s-[s])A} f_n(Y^n([s]), r([s])) \rangle_H \right. \\
 & \quad \left. + \| g_n(Y^n([s]), r([s])) \|_{HS}^2 \right\} ds \\
 & + \mathbb{E} \int_0^t e^{\theta s} \sum_{l=1}^N \gamma_{r(s)l} \lambda_l \| Y^n(s) \|_H^2 ds
 \end{aligned}$$

$$\begin{aligned}
 &\leq q\|x\|_H^2 + \theta q \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad - 2\alpha p \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \quad \times \left\{ 2 \langle Y^n(s), e^{(s-[s])A} f(Y^n([s]), r([s])) \rangle_H \right. \\
 &\quad \quad \quad \left. + \|g(Y^n([s]), r([s]))\|_{HS}^2 \right\} ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \sum_{l=1}^N \gamma_{r(s)l} \lambda_l \|Y^n(s)\|_H^2 ds.
 \end{aligned} \tag{31}$$

By the fundamental transformation, we obtain that

$$\begin{aligned}
 &\langle Y^n(t), e^{(t-[t])A} f(Y^n([t]), r([t])) \rangle_H \\
 &= \langle Y^n(t), f(Y^n(t), r(t)) \rangle_H \\
 &\quad + \langle Y^n(t), (e^{(t-[t])A} - \mathbf{1}) f(Y^n(t), r(t)) \rangle_H \\
 &\quad + \langle Y^n(t), e^{(t-[t])A} (f(Y^n([t]), r([t])) \\
 &\quad \quad \quad - f(Y^n(t), r(t))) \rangle_H.
 \end{aligned} \tag{32}$$

By Höld inequality, we have

$$\begin{aligned}
 &\|g(Y^n([t]), r([t]))\|_{HS}^2 \\
 &= \|g(Y^n(t), r(t)) \\
 &\quad - (g(Y^n(t), r(t)) - g(Y^n([t]), r([t])))\|_{HS}^2 \\
 &\leq (1 + \Delta^{1/2}) \|g(Y^n(t), r(t))\|_{HS}^2 + (1 + \Delta^{-1/2}) \\
 &\quad \times \|g(Y^n(t), r(t)) - g(Y^n([t]), r([t]))\|_{HS}^2.
 \end{aligned} \tag{33}$$

Then, from (31), we have

$$\begin{aligned}
 &e^{\theta t} \mathbb{E} (\lambda_{r(t)} \|Y^n(t)\|_H^2) \\
 &\leq q\|x\|_H^2 + \theta q \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad - 2\alpha p \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \left\{ 2 \langle Y^n(s), f(Y^n(s), r(s)) \rangle_H \right. \\
 &\quad \quad \left. + \|g(Y^n(s), r(s))\|_{HS}^2 \right\} ds
 \end{aligned}$$

$$\begin{aligned}
 &+ \mathbb{E} \int_0^t e^{\theta s} \sum_{l=1}^N \gamma_{r(s)l} \lambda_l \|Y^n(s)\|_H^2 ds \\
 &+ \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \times \left\{ 2 \langle Y^n(s), (e^{(s-[s])A} - \mathbf{1}) f(Y^n(s), r(s)) \rangle_H \right. \\
 &\quad \quad \left. + 2 \langle Y^n(s), e^{(s-[s])A} \right. \\
 &\quad \quad \quad \times (f(Y^n([s]), r([s])) \\
 &\quad \quad \quad \quad \left. - f(Y^n(s), r(s))) \rangle_H \right. \\
 &\quad \quad \left. + \Delta^{1/2} \|g(Y^n(s), r(s))\|_{HS}^2 + (1 + \Delta^{-1/2}) \right. \\
 &\quad \quad \left. \times \|g(Y^n(s), r(s)) \right. \\
 &\quad \quad \quad \left. - g(Y^n([s]), r([s]))\|_{HS}^2 \right\} ds
 \end{aligned}$$

$$\begin{aligned}
 &\leq q\|x\|_H^2 + \theta q \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad - 2\alpha p \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s)\|_H^2 ds \\
 &\quad - \frac{\mu}{2} \mathbb{E} \int_0^t e^{\theta s} \|Y(s)\|_H^2 ds + \alpha_1 \int_0^t e^{\theta s} ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \times \left\{ 2 \langle Y^n(s), (e^{(s-[s])A} - \mathbf{1}) f(Y^n(s), r(s)) \rangle_H \right. \\
 &\quad \quad \left. + 2 \langle Y^n(s), e^{(s-[s])A} (f(Y^n([s]), r([s])) \right. \\
 &\quad \quad \quad \left. - f(Y^n(s), r(s))) \rangle_H \right. \\
 &\quad \quad \left. + \Delta^{1/2} \|g(Y^n(s), r(s))\|_{HS}^2 + (1 + \Delta^{-1/2}) \right. \\
 &\quad \quad \left. \times \|g(Y^n(s), r(s)) \right. \\
 &\quad \quad \quad \left. - g(Y^n([s]), r([s]))\|_{HS}^2 \right\} ds \\
 &:= J_1(t) + J_2(t) + J_3(t) + J_4(t).
 \end{aligned} \tag{34}$$

By the elemental inequality: $2ab \leq (a^2/\kappa) + \kappa b^2$, $a, b \in \mathbb{R}$, $\kappa > 0$, and (8), (27), we obtain that, for $\Delta < 1$,

$$\begin{aligned}
 J_2(t) &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \left\{ \Delta^{1/2} \|Y^n(s)\|_H^2 \right. \\
 &\quad \left. + \Delta^{-1/2} \|(e^{(s-[s])A} - \mathbf{1}) \right. \\
 &\quad \quad \left. \times f(Y^n(s), r(s))\|_H^2 \right\} ds
 \end{aligned}$$

$$\begin{aligned}
 &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \Delta^{1/2} \|Y^n(s)\|_H^2 ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \Delta^{-1/2} \rho_n^2 \Delta^2 \bar{L} (1 + \|Y^n(s)\|_H^2) ds \\
 &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \{(\Delta^{1/2} + \Delta^{1/2} \rho_n^2 \bar{L}) \\
 &\quad \times \|Y^n(s)\|_H^2 + \Delta^{1/2} \rho_n^2 \bar{L}\} ds \\
 &\leq q \mathbb{E} \int_0^t e^{\theta s} \{\Delta^{1/2} (1 + \rho_n^2 \bar{L}) \|Y^n(s)\|_H^2 + \Delta^{1/2} \rho_n^2 \bar{L}\} ds.
 \end{aligned} \tag{35}$$

By (A2) and (8), we have

$$\begin{aligned}
 &\|(f(Y^n([t]), r([t])) - f(Y^n(t), r(t)))\|_H^2 \\
 &\leq 2\|(f(Y^n([t]), r([t])) - f(Y^n([t]), r(t)))\|_H^2 \\
 &\quad + 2\|(f(Y^n([t]), r(t)) - f(Y^n(t), r(t)))\|_H^2 \tag{36} \\
 &\leq 8\bar{L} (1 + \|Y^n([t])\|_H^2) I_{\{r(t) \neq r([t])\}} \\
 &\quad + 2L \|Y^n(t) - Y^n([t])\|_H^2.
 \end{aligned}$$

Similarly, we have

$$\begin{aligned}
 &\|g(Y^n(t), r(t)) - g(Y^n([t]), r([t]))\|_{HS}^2 \\
 &\leq 8\bar{L} (1 + \|Y^n([t])\|_H^2) I_{\{r(t) \neq r([t])\}} \\
 &\quad + 2L \|Y^n(t) - Y^n([t])\|_H^2.
 \end{aligned} \tag{37}$$

Thus, we obtain from (36) that

$$\begin{aligned}
 J_3(t) &\leq 2\mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \langle Y^n(s), e^{(s-[s])A} \\
 &\quad \times (f(Y^n([s]), r([s])) \\
 &\quad \quad - f(Y^n(s), r(s))) \rangle_H ds \\
 &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \left\{ \Delta^{1/2} \|Y^n(s)\|_H^2 ds + \Delta^{-1/2} \|e^{(s-[s])A}\| \right. \\
 &\quad \times \|f(Y^n([s]), r([s])) \\
 &\quad \quad \left. - f(Y^n(s), r(s))\|_H^2 \right\} ds \\
 &\leq q \mathbb{E} \int_0^t e^{\theta s} \left\{ \Delta^{1/2} \|Y^n(s)\|_H^2 + \Delta^{-1/2} 8\bar{L} \right. \\
 &\quad \times (1 + \|Y^n([s])\|_H^2) I_{\{r(s) \neq r([s])\}} \\
 &\quad \left. + 2\Delta^{-1/2} L \|Y^n(s) - Y^n([s])\|_H^2 \right\} ds.
 \end{aligned} \tag{38}$$

By Markov property, we compute

$$\begin{aligned}
 &\mathbb{E} \left[(1 + \|Y([t])\|_H^2) I_{\{r(t) \neq r([t])\}} \right] \\
 &= \mathbb{E} \left(\mathbb{E} \left[(1 + \|Y([t])\|_H^2) I_{\{r(t) \neq r([t])\}} \mid r([t]) \right] \right) \\
 &= \mathbb{E} \left(\mathbb{E} \left[(1 + \|Y([t])\|_H^2) \mid r([t]) \right] \right) \\
 &\quad \times \mathbb{E} \left[I_{\{r(t) \neq r([t])\}} \mid r([t]) \right] \\
 &= \mathbb{E} \left((1 + \|Y([t])\|_H^2) \sum_{i \in \mathbb{S}} I_{\{r([t])=i\}} \mathbb{P}(r(t) \neq i \mid r([t]) = i) \right) \\
 &= \mathbb{E} \left((1 + \|Y([t])\|_H^2) \sum_{i \in \mathbb{S}} I_{\{r([t])=i\}} \right) \\
 &\quad \times \sum_{j \neq i} (\gamma_{ij}(t - [t]) + o(t - [t])) \\
 &= \mathbb{E} \left((1 + \|Y([t])\|_H^2) \left(\max_{i \in \mathbb{S}} (-\gamma_{ii}) \Delta + o(\Delta) \right) \sum_{i \in \mathbb{S}} I_{\{r([t])=i\}} \right) \\
 &\leq \hat{\gamma} \Delta \mathbb{E} (1 + \|Y([t])\|_H^2),
 \end{aligned} \tag{39}$$

where $\hat{\gamma} = N[1 + \max_{1 \leq i \leq N} (-\gamma_{ii})]$. Substituting (39) into (38) gives

$$\begin{aligned}
 J_3(t) &\leq q \mathbb{E} \int_0^t e^{\theta s} \left\{ \Delta^{1/2} \|Y^n(s)\|_H^2 \right. \\
 &\quad \left. + 2\Delta^{-1/2} L \|Y^n(s) - Y^n([s])\|_H^2 \right\} ds \\
 &\quad + q \int_0^t 8e^{\theta s} \Delta^{1/2} \hat{\gamma} \bar{L} \mathbb{E} (1 + \|Y([s])\|_H^2) ds.
 \end{aligned} \tag{40}$$

Furthermore, due to (37) and (39), we have

$$\begin{aligned}
 J_4(t) &= \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \times \left\{ \Delta^{1/2} \|g(Y^n(s), r(s))\|_{HS}^2 \right. \\
 &\quad \left. + (1 + \Delta^{-1/2}) \|g(Y^n(s), r(s)) \right. \\
 &\quad \quad \left. - g(Y^n([s]), r([s]))\|_{HS}^2 \right\} ds
 \end{aligned}$$

$$\begin{aligned}
 &\leq q\mathbb{E} \int_0^t e^{\theta s} \Delta^{1/2} \bar{L} (1 + \|Y^n(s)\|_H^2) ds + q(1 + \Delta^{-1/2}) \\
 &\quad \times \int_0^t e^{\theta s} 8\bar{L} (1 + \|Y^n(\lfloor s \rfloor)\|_H^2) I_{\{r(s) \neq r(\lfloor s \rfloor)\}} ds \\
 &\quad + 2Lq(1 + \Delta^{-1/2}) \int_0^t e^{\theta s} \|Y^n(s) - Y^n(\lfloor s \rfloor)\|_H^2 ds \\
 &\leq q\Delta^{1/2} \bar{L} \mathbb{E} \int_0^t e^{\theta s} (1 + \|Y^n(s)\|_H^2) ds \\
 &\quad + 16q\hat{\gamma}\Delta^{1/2} \bar{L} \int_0^t e^{\theta s} (1 + \|Y^n(s)\|_H^2) ds \\
 &\quad + 2Lq(1 + \Delta^{-1/2}) \int_0^t e^{\theta s} \|Y^n(s) - Y^n(\lfloor s \rfloor)\|_H^2 ds.
 \end{aligned} \tag{41}$$

On the other hand, by Lemma 10, when $3(\rho_n^2 + 2\bar{L})\Delta \leq 1$, we have

$$\begin{aligned}
 &\mathbb{E} \|Y^n(t)\|_H^2 \\
 &\leq 2\mathbb{E} \|Y^n(t) - Y^n(\lfloor t \rfloor)\|_H^2 + 2\mathbb{E} \|Y^n(\lfloor t \rfloor)\|_H^2 \\
 &\leq 6(\rho_n^2 + 2\bar{L})\Delta (1 + \mathbb{E} \|Y^n(\lfloor t \rfloor)\|_H^2) \\
 &\quad + 2\mathbb{E} \|Y^n(\lfloor t \rfloor)\|_H^2 \\
 &\leq 4\mathbb{E} \|Y^n(\lfloor t \rfloor)\|_H^2 + 2.
 \end{aligned} \tag{42}$$

Putting (35), (40), and (41) into (34), we have

$$\begin{aligned}
 &e^{\theta t} \mathbb{E} (\lambda_{r(t)} \|Y^n(t)\|_H^2) \\
 &\leq q\|x\|_H^2 + \int_0^t e^{\theta s} [\alpha_1 + q\rho_n^2 \Delta^{1/2} \bar{L} \\
 &\quad + 24q\Delta^{1/2} \hat{\gamma} \bar{L} + q\Delta^{1/2} \bar{L}] ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \left[q\theta - 2\alpha p - \frac{\mu}{2} + q\Delta^{1/2} (2 + \rho_n^2 \bar{L}) + q\Delta^{1/2} \bar{L} \right] \\
 &\quad \times \|Y^n(s)\|_H^2 ds + 24q\Delta^{1/2} \hat{\gamma} \bar{L} \mathbb{E} \\
 &\quad \times \int_0^t e^{\theta s} \|Y^n(\lfloor s \rfloor)\|_H^2 ds \\
 &\quad + (4q\Delta^{-1/2} L + 2qL) \mathbb{E} \int_0^t e^{\theta s} \|Y^n(s) - Y^n(\lfloor s \rfloor)\|_H^2 ds.
 \end{aligned} \tag{43}$$

By Lemma 10 and the inequality (42), we obtain that

$$\begin{aligned}
 &e^{\theta t} \mathbb{E} (\lambda_{r(t)} \|Y^n(t)\|_H^2) \\
 &\leq q\|x\|_H^2 + \int_0^t e^{\theta s} [\alpha_1 + 2q\theta - 4\alpha p - \mu \\
 &\quad + 3q\rho_n^2 \Delta^{1/2} \bar{L} + 24q\Delta^{1/2} \hat{\gamma} \bar{L} \\
 &\quad + 3q\Delta^{1/2} \bar{L} + 4q\Delta^{1/2}] ds \\
 &\quad + \int_0^t e^{\theta s} [4q\theta - 8\alpha p - 2\mu + 4q\Delta^{1/2} (2 + \rho_n^2 \bar{L}) \\
 &\quad + 4q\Delta^{1/2} \bar{L} + 24q\Delta^{1/2} \hat{\gamma} \bar{L}] \|Y^n(\lfloor s \rfloor)\|_H^2 ds \\
 &\quad + 6qL\Delta^{1/2} (\rho_n^2 + 2\bar{L}) \mathbb{E} \int_0^t e^{\theta s} (1 + \|Y^n(\lfloor s \rfloor)\|_H^2) ds \\
 &\leq q\|x\|_H^2 + \int_0^t e^{\theta s} [\alpha_1 + 2q\theta - 4\alpha p - \mu + 3q\rho_n^2 \Delta^{1/2} \bar{L} \\
 &\quad + 24q\Delta^{1/2} \hat{\gamma} \bar{L} + 3q\Delta^{1/2} \bar{L} + 4q\Delta^{1/2} \\
 &\quad + 6qL\Delta^{1/2} (\rho_n^2 + 2\bar{L})] ds \\
 &\quad + \int_0^t e^{\theta s} [4q\theta - 8\alpha p - 2\mu + 4q\Delta^{1/2} (2 + \rho_n^2 \bar{L}) \\
 &\quad + 4q\Delta^{1/2} \bar{L} + 24q\Delta^{1/2} \hat{\gamma} \bar{L} \\
 &\quad + 6qL\Delta^{1/2} (\rho_n^2 + 2\bar{L})] \|Y^n(\lfloor s \rfloor)\|_H^2 ds.
 \end{aligned} \tag{44}$$

Let $\theta = (4\alpha p + \mu)/4q$, for $\Delta < ((4\alpha p + \mu)/(8q + 4q\rho_n^2 \bar{L} + 4q\bar{L} + 24q\hat{\gamma} + 6qL(\rho_n^2 + 2\bar{L})))^2$, then

$$p e^{\theta t} \mathbb{E} (\|Y(t)\|_H^2) \leq q\|x\|_H^2 + \int_0^t e^{\theta s} \left[\alpha_1 + \frac{4\alpha p + \mu}{2} \right] ds. \tag{45}$$

That is,

$$\sup_{t \geq 0} \mathbb{E} (\|Y(t)\|_H^2) \leq C. \tag{46}$$

□

Lemma 12. Let (A1)–(A3) hold. If $\Delta < \min\{1, 1/18(\rho_n^2 + 2L), ((2\alpha p + \mu)/(4q + 2qL + 2q\rho_n^2 L + 12qL\hat{\gamma}))^2\}$, then

$$\lim_{t \rightarrow \infty} \mathbb{E} \|Y^{n,x,i}(t) - Y^{n,y,i}(t)\|_H^2 = 0 \quad \text{uniformly for } x, y \in U, \tag{47}$$

where U is a bounded subset of H_n .

Proof. Write $Y^{n,x,i}(t) = Y^x(t)$, $Y^{n,y,i}(t) = Y^y(t)$, $r^i(k\Delta) = r(k\Delta)$. From (20), it is easy to show that

$$\begin{aligned} & (Y^x(t) - Y^y(t)) - (Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor)) \\ &= (Y^x(t) - Y^x(\lfloor t \rfloor)) - (Y^y(t) - Y^y(\lfloor t \rfloor)) \\ &= (e^{(t-\lfloor t \rfloor)A} - \mathbf{1})(Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor)) \\ &+ \int_{\lfloor t \rfloor}^t e^{(t-s)A} (f_n(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ &\quad - f_n(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))) ds \\ &+ \int_{\lfloor t \rfloor}^t e^{(t-s)A} (g_n(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ &\quad - g_n(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))) dW(s). \end{aligned} \tag{48}$$

By using the argument of Lemma 10, we derive that, if $\Delta < 1$,

$$\begin{aligned} & \mathbb{E} \|(Y^x(t) - Y^y(t)) - (Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2 \\ & \leq 3(\rho_n^2 + 2L) \Delta \mathbb{E} \|Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor)\|_H^2 \\ & \mathbb{E} \|(Y^x(t) - Y^y(t))\|_H^2 \\ &= \mathbb{E} \|(Y^x(t) - Y^y(t)) - (Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor)) \\ &\quad + (Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2 \\ & \leq (1 + 2) \mathbb{E} \|(Y^x(t) - Y^y(t)) \\ &\quad - (Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2 \\ &\quad + \left(1 + \frac{1}{2}\right) \mathbb{E} \|Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor)\|_H^2 \\ & \leq 9(\rho_n^2 + 2L) \Delta \mathbb{E} \|(Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2 \\ &\quad + 1.5 \mathbb{E} \|(Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2. \end{aligned} \tag{49}$$

If $\Delta < 1/18(\rho_n^2 + 2L)$, then

$$\mathbb{E} \|(Y^x(t) - Y^y(t))\|_H^2 \leq 2 \mathbb{E} \|(Y^x(\lfloor t \rfloor) - Y^y(\lfloor t \rfloor))\|_H^2. \tag{51}$$

Using (30) and the generalised Itô formula, for any $\theta > 0$, we have

$$\begin{aligned} & e^{\theta t} \mathbb{E} (\lambda_{r(t)} \|Y^x(t) - Y^y(t)\|_H^2) \\ & \leq \lambda_i \|x - y\|_H^2 \\ & + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \{ \theta \|Y^x(s) - Y^y(s)\|_H^2 \\ & \quad + 2 \langle Y^x(s) - Y^y(s), \\ & \quad AY^x(s) - Y^y(s) \rangle_H \end{aligned}$$

$$\begin{aligned} & + 2 \langle Y^x(s) - Y^y(s), e^{(s-\lfloor s \rfloor)A} \\ & \quad \times (f_n(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ & \quad - f_n(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))) \rangle_H \\ & + \|g_n(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ & \quad - g_n(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))\|_{HS}^2 \} ds \\ & + \mathbb{E} \int_0^t e^{\theta s} \sum_{l=1}^N \gamma_{r(s)l} \lambda_l \|Y^x(s) - Y^y(s)\|_H^2 ds \\ & \leq q \|x - y\|_H^2 + q\theta \mathbb{E} \int_0^t e^{\theta s} \|Y^x(s) - Y^y(s)\|_H^2 ds \\ & \quad - 2\alpha p \mathbb{E} \int_0^t e^{\theta s} \|Y^x(s) - Y^y(s)\|_H^2 ds \\ & + \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \{ 2 \langle Y^x(s) - Y^y(s), e^{(s-\lfloor s \rfloor)A} \\ & \quad \times (f(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ & \quad - f(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))) \rangle_H \\ & \quad + \|g(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\ & \quad - g(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))\|_{HS}^2 \} ds \\ & + \mathbb{E} \int_0^t e^{\theta s} \sum_{l=1}^N \gamma_{r(s)l} \lambda_l \|Y^x(s) - Y^y(s)\|_H^2 ds. \end{aligned} \tag{52}$$

By the fundamental transformation, we obtain that

$$\begin{aligned} & \langle Y^x(t) - Y^y(t), e^{(t-\lfloor t \rfloor)A} \\ & \quad \times (f(Y^x(\lfloor t \rfloor), r(\lfloor t \rfloor)) - f(Y^y(\lfloor t \rfloor), r(\lfloor t \rfloor))) \rangle_H \\ &= \langle Y^x(t) - Y^y(t), f(Y^x(t), r(t)) - f(Y^y(t), r(t)) \rangle_H \\ & + \langle Y^x(t) - Y^y(t), (e^{(t-\lfloor t \rfloor)A} - \mathbf{1}) \\ & \quad \times (f(Y^x(\lfloor t \rfloor), r(\lfloor t \rfloor)) - f(Y^y(\lfloor t \rfloor), r(\lfloor t \rfloor))) \rangle_H \\ & + \langle Y^x(t) - Y^y(t), e^{(t-\lfloor t \rfloor)A} \\ & \quad \times (f(Y^x(\lfloor t \rfloor), r(\lfloor t \rfloor)) - f(Y^y(\lfloor t \rfloor), r(\lfloor t \rfloor))) \\ & \quad - (f(Y^x(t), r(t)) - f(Y^y(t), r(t))) \rangle_H. \end{aligned} \tag{53}$$

By the Höld inequality, we have

$$\begin{aligned} & \|g(Y^x(\lfloor t \rfloor), r(\lfloor t \rfloor)) - g(Y^y(\lfloor t \rfloor), r(\lfloor t \rfloor))\|_{HS}^2 \\ & \leq (1 + \Delta^{1/2}) \|g(Y^x(t), r(t)) - g(Y^y(t), r(t))\|_{HS}^2 \end{aligned}$$

$$\begin{aligned}
 &+ (1 + \Delta^{-1/2}) \|(g(Y^x(t), r(t)) - g(Y^y(t), r(t))) \\
 &\quad - (g(Y^x(\lfloor t \rfloor), r(\lfloor t \rfloor)) \\
 &\quad - g(Y^y(\lfloor t \rfloor), r(\lfloor t \rfloor)))\|_{HS}^2. \tag{54}
 \end{aligned}$$

Then, from (52) and (A3), we have

$$\begin{aligned}
 &e^{\theta t} \mathbb{E} (\lambda_{r(t)} \|Y^x(t) - Y^y(t)\|_H^2) \\
 &\leq q \|x - y\|_H^2 + (q\theta - 2\alpha p - \mu) \mathbb{E} \\
 &\quad \times \int_0^t e^{\theta s} \|Y^x(s) - Y^y(s)\|_H^2 ds \\
 &+ 2\mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \times \langle Y^x(s) - Y^y(s), (e^{(s-\lfloor s \rfloor)A} - \mathbf{1}) \\
 &\quad \times (f(Y^x(s), r(s)) \\
 &\quad - f(Y^y(s), r(s))) \rangle_H ds \\
 &+ 2\mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \\
 &\quad \times \langle Y^x(s) - Y^y(s), e^{(s-\lfloor s \rfloor)A} \\
 &\quad \times (f(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\
 &\quad - f(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))) \\
 &\quad - (f(Y^x(s), r(s)) \\
 &\quad - f(Y^y(s), r(s))) \rangle_H ds \\
 &+ \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \{ \Delta^{1/2} \|g(Y^x(s), r(s)) \\
 &\quad - g(Y^y(s), r(s))\|_{HS}^2 \\
 &\quad + (1 + \Delta^{-1/2}) \\
 &\quad \times \|(g(Y^x(s), r(s)) \\
 &\quad - g(Y^y(s), r(s))) \\
 &\quad - (g(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\
 &\quad - g(Y^y(\lfloor s \rfloor), \\
 &\quad r(\lfloor s \rfloor)))\|_{HS}^2 \} ds \\
 &:= G_1(t) + G_2(t) + G_3(t) + G_4(t).
 \end{aligned} \tag{55}$$

By (A2) and (27), we have, for $\Delta < 1$,

$$\begin{aligned}
 &G_2(t) \\
 &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} \{ \Delta^{1/2} \|Y^x(s) - Y^y(s)\|_H^2 \\
 &\quad + \Delta^{-1/2} \|(e^{(s-\lfloor s \rfloor)A} - \mathbf{1})
 \end{aligned}$$

$$\begin{aligned}
 &\times (f(Y^x(s), r(s)) \\
 &\quad - Y^y(s), r(s))\|_H^2 \} ds \\
 &\leq \mathbb{E} \int_0^t e^{\theta s} \lambda_{r(s)} (\Delta^{1/2} + \Delta^{3/2} \rho_n^2 L) \|Y^x(s) - Y^y(s)\|_H^2 ds \\
 &\leq q \mathbb{E} \int_0^t e^{\theta s} \Delta^{1/2} (1 + \rho_n^2 L) \|Y^x(s) - Y^y(s)\|_H^2 ds. \tag{56}
 \end{aligned}$$

It is easy to show that

$$\begin{aligned}
 &G_3(t) \\
 &\leq q \mathbb{E} \int_0^t e^{\theta s} \{ \Delta^{1/2} \|Y^x(s) - Y^y(s)\|_H^2 \\
 &\quad + \Delta^{-1/2} \|(e^{(s-\lfloor s \rfloor)A})\|^2 \\
 &\quad \times \|f(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\
 &\quad - f(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\
 &\quad - (f(Y^x(s), r(s)) \\
 &\quad - Y^y(s), r(s))\|_H^2 \} ds \\
 &\leq q \mathbb{E} \int_0^t e^{\theta s} \Delta^{1/2} \|Y^x(s) - Y^y(s)\|_H^2 ds + \bar{G}_3(t).
 \end{aligned} \tag{57}$$

By (39), we have

$$\begin{aligned}
 &\bar{G}_3(t) \\
 &\leq 2q \Delta^{-1/2} \mathbb{E} \int_0^t e^{\theta s} [\|f(Y^x(\lfloor s \rfloor), r(\lfloor s \rfloor)) \\
 &\quad - f(Y^y(\lfloor s \rfloor), r(\lfloor s \rfloor))\|_H^2 \\
 &\quad + \|(f(Y^x(s), r(s)) \\
 &\quad - Y^y(s), r(s))\|_H^2] \\
 &\quad \times I_{\{r(s) \neq r(\lfloor s \rfloor)\}} ds \\
 &\leq 4q \Delta^{-1/2} L \mathbb{E} \int_0^t e^{\theta s} \|Y^x(\lfloor s \rfloor) - Y^y(\lfloor s \rfloor)\|_H^2 \\
 &\quad \times I_{\{r(s) \neq r(\lfloor s \rfloor)\}} ds \\
 &\leq 4qL\hat{\gamma} \Delta^{1/2} \mathbb{E} \int_0^t e^{\theta s} \|Y^x(\lfloor s \rfloor) - Y^y(\lfloor s \rfloor)\|_H^2 ds.
 \end{aligned} \tag{58}$$

Therefore, we obtain that

$$\begin{aligned}
 &G_3(t) \leq q \Delta^{1/2} \mathbb{E} \int_0^t e^{\theta s} \|Y^x(s) - Y^y(s)\|_H^2 ds \\
 &\quad + 4qL\hat{\gamma} \Delta^{1/2} \mathbb{E} \int_0^t e^{\theta s} \|Y^x(\lfloor s \rfloor) - Y^y(\lfloor s \rfloor)\|_H^2 ds. \tag{59}
 \end{aligned}$$

On the other hand, using the similar argument of (58), we have

$$\begin{aligned}
 G_4(t) &\leq q\mathbb{E} \int_0^t e^{\theta s} \left\{ \Delta^{1/2} L \|Y^x(s) - Y^y(s)\|_H^2 \right. \\
 &\quad \left. + (1 + \Delta^{-1/2}) 4L\widehat{\gamma}\Delta \right. \\
 &\quad \left. \times \|Y^x([s]) - Y^y([s])\|_H^2 \right\} ds. \tag{60}
 \end{aligned}$$

Hence, we have

$$\begin{aligned}
 &pe^{\theta t} \mathbb{E} \left(\|Y^x(t) - Y^y(t)\|_H^2 \right) \\
 &\leq q\|x - y\|_H^2 \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \left[q\theta - 2\alpha p - \mu + q(1 + \rho_n^2 L) \Delta^{1/2} \right. \\
 &\quad \left. + q\Delta^{1/2} + qL\Delta^{1/2} \right] \|Y^x(s) - Y^y(s)\|_H^2 ds \\
 &\quad + \mathbb{E} \int_0^t e^{\theta s} \left[4qL\widehat{\gamma}\Delta^{1/2} + (\Delta^{1/2} + \Delta) 4qL\widehat{\gamma} \right] \\
 &\quad \times \|Y^x([s]) - Y^y([s])\|_H^2 ds. \tag{61}
 \end{aligned}$$

By (50), we obtain that

$$\begin{aligned}
 &pe^{\theta t} \mathbb{E} \left(\|Y^x(t) - Y^y(t)\|_H^2 \right) \\
 &\leq q\|x - y\|_H^2 + \mathbb{E} \int_0^t e^{\theta s} \left[2q\theta - 4\alpha p - 2\mu \right. \\
 &\quad \left. + 4q\Delta^{1/2} + 2qL\Delta^{1/2} \right. \\
 &\quad \left. + 2q\rho_n^2 L\Delta^{1/2} + 12qL\widehat{\gamma}\Delta^{1/2} \right] \\
 &\quad \times \|Y^x([s]) - Y^y([s])\|_H^2 ds. \tag{62}
 \end{aligned}$$

Let $\theta = (2\alpha p + \mu)/2q$, for $\Delta < ((2\alpha p + \mu)/(4q + 2qL + 2q\rho_n^2 L + 12qL\widehat{\gamma}))^2$, then the desired assertion (47) follows. \square

We can now easily prove our main result.

Proof of Theorem 9. Since H_n is finite-dimensional, by Lemma 3.1 in [12], we have

$$\lim_{k \rightarrow \infty} d_{\perp} \left(\mathbb{P}_k^{n,\Delta}((x, i), \cdot \times \cdot), \mathbb{P}_k^{n,\Delta}((y, i), \cdot \times \cdot) \right) = 0, \tag{63}$$

uniformly in $x, y \in H_n, i, j \in \mathbb{S}$.

By Lemma 7, there exists $\pi^{n,\Delta}(\cdot \times \cdot) \in \mathcal{P}(H_n \times \mathbb{S})$, such that

$$\lim_{k \rightarrow \infty} d_{\perp} \left(\mathbb{P}_k^{n,\Delta}((0, 1), \cdot \times \cdot), \pi^{n,\Delta}(\cdot \times \cdot) \right) = 0. \tag{64}$$

By the triangle inequality (63) and (64), we have

$$\begin{aligned}
 &\lim_{k \rightarrow \infty} d_{\perp} \left(\mathbb{P}_k^{n,\Delta}((x, i), \cdot \times \cdot), \pi^{n,\Delta}(\cdot \times \cdot) \right) \\
 &\leq \lim_{k \rightarrow \infty} d_{\perp} \left(\mathbb{P}_k^{n,\Delta}((x, i), \cdot \times \cdot), \mathbb{P}_k^{n,\Delta}((0, 1), \cdot \times \cdot) \right) \\
 &\quad + \lim_{k \rightarrow \infty} d_{\perp} \left(\mathbb{P}_k^{n,\Delta}((0, 1), \cdot \times \cdot), \pi^{n,\Delta}(\cdot \times \cdot) \right) = 0.
 \end{aligned} \tag{65}$$

\square

4. Corollary and Example

In this section, we give a criterion based M -matrices which can be verified easily in applications.

(A4) For each $j \in \mathbb{S}$, there exists a pair of constants β_j and δ_j such that, for $x, y \in H$,

$$\begin{aligned}
 \langle x - y, f(x, j) - f(y, j) \rangle_H &\leq \beta_j \|x - y\|_H^2, \\
 \|g(x, j) - g(y, j)\|_{HS}^2 &\leq \delta_j \|x - y\|_H^2.
 \end{aligned} \tag{66}$$

Moreover, $\mathcal{A} := -\text{diag}(2\beta_1 + \delta_1, \dots, 2\beta_N + \delta_N) - \Gamma$ is a nonsingular M -matrix [8].

Corollary 13. Under (A1), (A2), and (A4), for a given stepsize $\Delta > 0$, and arbitrary $x \in H_n, i \in \mathbb{S}$, $\{\bar{Z}^{n,(x,i)}(k\Delta)\}_{k \geq 0}$ has a unique stationary distribution $\pi^{n,\Delta}(\cdot \times \cdot) \in \mathcal{P}(H_n \times \mathbb{S})$.

Proof. In fact, we only need to prove that (A3) holds. By (A4), there exists $(\lambda_1, \lambda_2, \dots, \lambda_N)^T > 0$, such that $(q_1, q_2, \dots, q_N)^T = \mathcal{A}(\lambda_1, \lambda_2, \dots, \lambda_N)^T > 0$.

Set $\mu = \min_{1 \leq j \leq N} q_j$, by (66), we have

$$\begin{aligned}
 &2\lambda_j \langle x - y, f(x, j) - f(y, j) \rangle_H \\
 &\quad + \lambda_j \|g(x, j) - g(y, j)\|_H^2 + \sum_{l=1}^N \gamma_{jl} \lambda_l \|x - y\|_H^2 \\
 &\leq 2\lambda_j \beta_j \|x - y\|_H^2 + \delta_j \lambda_j \|x - y\|_H^2 + \sum_{l=1}^N \gamma_{jl} \lambda_l \|x - y\|_H^2 \\
 &= \left(2\lambda_j \beta_j + \delta_j \lambda_j + \sum_{l=1}^N \gamma_{jl} \lambda_l \right) \|x - y\|_H^2 \\
 &= -q_j \|x - y\|_H^2 \leq -\mu \|x - y\|_H^2
 \end{aligned} \tag{67}$$

\square

In the following, we give an example to illustrate the Corollary 13.

Example 14. Consider

$$\begin{aligned}
 dX(t, \xi) &= \left[\frac{\partial^2}{\partial \xi^2} X(t, \xi) + B(r(t)) X(t, \xi) \right] dt \\
 &\quad + g(X(t, \xi), r(t)) dW(t), \quad 0 < \xi < \pi, t \geq 0.
 \end{aligned} \tag{68}$$

We take $H = L^2(0, \pi)$ and $A = \partial^2/\partial\xi^2$ with domain $\mathcal{D}(A) = H^2(0, \pi) \cap H_0^1(0, \pi)$, then A is a self-adjoint negative operator. For the eigenbasis $e_k(\xi) = (2/\pi)^{1/2} \sin(k\xi)$, $\xi \in [0, \pi]$, $Ae_k = -k^2e_k$, $k \in \mathbb{N}$. It is easy to show that

$$\|e^{tA}x\|_H^2 = \sum_{i=1}^{\infty} e^{-2k^2t} \langle x, e_i \rangle_H^2 \leq e^{-2t} \sum_{i=1}^{\infty} \langle x, e_i \rangle_H^2. \quad (69)$$

This further gives that

$$\|e^{tA}\| \leq e^{-t}, \quad (70)$$

where $\alpha = 1$, thus (A1) holds.

Let $W(t)$ be a scalar Brownian motion, let $r(t)$ be a continuous-time Markov chain values in $\mathbb{S} = 1, 2$, with the generator

$$\begin{aligned} \Gamma &= \begin{pmatrix} -2 & 2 \\ 1 & -1 \end{pmatrix}, \\ B(1) = B_1 &= \begin{pmatrix} -0.3 & -0.1 \\ -0.2 & -0.2 \end{pmatrix}, \\ B(2) = B_2 &= \begin{pmatrix} -0.4 & -0.2 \\ -0.3 & -0.2 \end{pmatrix}. \end{aligned} \quad (71)$$

Then $\lambda_{\max}(B_1^T B_1) = 0.1706$, $\lambda_{\max}(B_2^T B_2) = 0.3286$.

Moreover, g satisfies

$$\|g(x, j) - g(y, j)\|_{HS}^2 \leq \delta_j \|x - y\|_H^2, \quad (72)$$

where $\delta_1 = 0.1$, $\delta_2 = 0.06$.

Defining $f(x, j) = B(j)x$, then

$$\begin{aligned} &\|f(x, j) - f(y, j)\|_{HS}^2 \vee \|g(x, j) - g(y, j)\|_{HS}^2 \\ &\leq (\lambda_{\max}(B_j^T B_j) \vee \delta_j) \|x - y\|_H^2 < 0.33 \|x - y\|_H^2 \\ &\quad \langle x - y, f(x, j) - f(y, j) \rangle_H \\ &\leq \frac{1}{2} \langle x - y, (B_j^T + B_j)(x - y) \rangle_H \\ &\leq \frac{1}{2} \lambda_{\max}(B_j^T + B_j) \|x - y\|_H^2. \end{aligned} \quad (73)$$

It is easy to compute

$$\begin{aligned} \beta_1 &= \frac{1}{2} \lambda_{\max}(B_1^T + B_1) = -0.0919, \\ \beta_2 &= \frac{1}{2} \lambda_{\max}(B_2^T + B_2) = -0.03075. \end{aligned} \quad (74)$$

So the matrix \mathcal{A} becomes

$$\mathcal{A} = \text{diag}(0.0838, 0.0015) - \Gamma = \begin{pmatrix} 2.0838 & -2 \\ -1 & 1.0015 \end{pmatrix}. \quad (75)$$

It is easy to see that \mathcal{A} is a nonsingular M -matrix. Thus, (A4) holds. By Corollary 13, we can conclude that (68) has a unique stationary distribution $\pi^{n,\Delta}(\cdot \times \cdot)$.

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