Mean-square stability of the zero equilibrium of the nonlinear delay differential equation: Nicholson’s blowflies application

H. El-Metwally · M. A. Sohaly · I. M. Elbaz

Abstract We are concerned about the stochastic nonlinear delay differential equation. The stochasticity arises from the white Gaussian noise, which is the time derivative of the standard Brownian motion. The main objective of this paper is to introduce a new technique using the Lyapunov functional for the study of stability of the zero solution of the stochastic delay differential system. Constructing a new appropriate deterministic system in the neighborhood of the origin is an effective way to investigate the necessary and sufficient conditions of stability in the sense of the mean square. Nicholson’s blowflies equation is one of the major problems in ecology; necessary conditions for the possible extinction of the Nicholson’s blowflies population are investigated. We support our theoretical results by providing areas of stability and some numerical simulations of the solution of the system using the Euler–Maruyama scheme, which is mean square stable Maruyama (Rendiconti del Circolo Matematico di Palermo 4(1):48, 1955), Cao et al. (Appl Math Comput 159(1):127–135, 2004).

Keywords Lyapunov functional · Mean-square stability · Zero equilibrium · Nicholson’s blowflies model

1 Introduction

Functional differential equations (FDEs) have motivated many mathematical and applied statistical researches. For instance, one of the major problems is the dynamics of animal population that has been proposed by [3–6]. Time delay occurs so often; many processes involve time delays in physics, medicine, finance, ecology, chemistry, etc. Realistic models must include some of the past history of the state of the system, and this, in turn, leads to the delay differential equations (DDEs). For more details regarding processes with time delay and their qualitative behavior, see [3–5, 7–12].

In these epidemiological and ecological mathematical models, uncertainty in the interactions between individuals of population and/or higher-frequency environmental noises develops stochasticity. Considering the stochasticity in these models reveals the mechanisms that influence the transmission and control the disease in epidemiology, dynamics of the population in ecology, etc. [13]. Stochastic models are proposed to capture the uncertainty and variations in the mathematical models by perturbing the deterministic system...
with a white Gaussian noise, which is ill-defined, and then change it to a Brownian motion, which is well defined by
\[ \zeta(t) dt = dB(t). \]

Without solving these systems, we study the stability of the equilibria, which is a very effective way to have a good insight of the solutions and their properties. Constructing a suitable Lyapunov functional is a good approach for investigating the necessary and sufficient conditions of stability of the equilibrium points [10,14–16].

Dedicated to the study of stability of stochastic delay differential equations (SDDEs), there are many works, including, but not limited to, use this class of equations to control the stabilization problem of the controlled inverted pendulum, model the infectious diseases and model the population dynamics of the Australian sheep-blowfly known as Nicholson’s blowfly [17–19]. By stochastic neutral differential equations (SNDEs), the stochastic neural networks and stochastic cellular neural networks have been used to model many of the human activities in science [20–22].

Our work is involved in the study of stability of delayed dynamical systems. In comparison with the known methods of investigating stability in the mean square of these systems [10,18,19,23], our approach is more effective as it lays in constructing an appropriate deterministic system using appropriate Lyapunov functionals. It can be implemented in many application problems, see [4,20,22]. For instance, regarding the Nicholson’s blowflies model, many works in the literature have studied the extinction of these blowflies in the deterministic and stochastic sense [18,24–26]. Our concept provides more stability delay-dependent and delay-independent conditions with better stability regions compared to the work done on this equation. Moreover, by this approach, this work can be extended by studying the persistence of these species of blowflies.

We shall determine the mean-square stability conditions of the stochastic nonlinear delay differential equation and its general form using a new way of constructing a delayed-deterministic system by Lyapunov functional in the presence of the white noise term. Consider the stochastic delay differential equation
\[
\begin{aligned}
&\begin{cases}
\frac{dX(t)}{dt} = \mu(t, X_t) dt + \sigma(t, X_t) dB(t), \\
X_t := X(t+s), & -\tau < s \leq 0 \\
X(0) = \phi(s), & \phi \in \mathcal{C}^+.
\end{cases}
\end{aligned}
\]

The solution of this equation is
\[ X(t) = \phi(0) + \int_0^t \mu(s, X_s) ds + \int_0^t \sigma(s, X_s) dB(s), \]

where the last term is known as an Itô integral. Existence and uniqueness of this solution are considered in [27,28]. The delay in the stochastic process is \( \tau > 0 \), assume \( C := C([-\tau, 0], \mathbb{R}) \) is a Banach space of continuous functions defined on \([-\tau, 0]\) with the norm \( \|\phi\| = \sup_{-\tau < s \leq 0} |\phi(s)| \). In population dynamics, let the initial function \( \phi \in \mathcal{C}^+ \), where \( \mathcal{C}^+ := C([-\tau, 0], \mathbb{R}^+) \). The continuous functionals \( \mu(t, X_t) \) and \( \sigma(t, X_t) \) are defined on \([0, \infty) \times \mathcal{C}[-\tau, 0] \) and satisfy Lipschitz condition in the second argument, i.e., for \( L > 0 \), \[ |\mu_1(t, X_t) - \mu_2(t, X^*_t)| \leq L |X_t - X^*_t|, \]

where \( \mathcal{C}^+ \) is the filtration generated by it up to time \( t \). The initial function \( \phi \) is a stochastic process which is independent of the minimal \( \sigma \)-algebra generated by random variables \( B(t) - B(s) \) for \( 0 \leq s < t < \infty \). Moreover, we also consider the general stochastic delay differential equation
\[
\frac{dX(t)}{dt} = \mu(t, X_t) dt + \sum_{i=1}^m \sigma_i(t, X_t) dB_i(t),
\]

where the functionals \( \mu(t, X_t) \) and \( \sigma(t, X_t) \) are \( m \)-dimensional continuously differentiable. \( B_i(t), \ i = 1, \cdots, m \) are \( m \)-dimensional standard Wiener processes.

1.1 Nicholson’s blowflies model

One of the major problems in ecology is the population dynamics of the Australian sheep blowfly, Lucilia cuprina, which is known as Nicholson’s blowfly. Author in [29] introduced the differential equation that models the population dynamics of this blowfly with fixed delay in the following form
\[
\frac{dX(t)}{dt} = pX(t-\tau) e^{-aX(t-\tau)} - \delta X(t).
\]
$X(t)$ represents the population of the mature adults at time $t$. All parameters $p$, $a$, $\delta \in [0, \infty)$, where $p$ represents the maximum daily production rate of eggs per capita, $a^{-1}$ is the size at which the population reproduces at the maximum rate, $\delta$ is the adult death rate of per capita daily and $\tau > 0$ is the delay in the production process.

The basic reproduction ratio is denoted by
$$R_0 = \frac{p}{\delta}.$$ 

This model admits only the trivial solution, $X_0 = 0$, if the basic reproduction ratio of the system $R_0 < 1$, i.e., the maximum reproduction rate per capita, is less than the death rate, and for $R_0 > 1$, a unique positive equilibrium, $X_* = \frac{1}{a} \ln \left( \frac{p}{\delta} \right)$, exists and at this point, this kind of blowflies does not become extinct. The species become extinct due to the global stability of the zero solution of (4), see [30]. Many authors have studied the dynamics of this equation in the deterministic sense [23, 26, 30–32], and stochastically [18, 19, 24, 25, 33].

This system is exposed to stochastic perturbation in the form of white noise, which is assumed to be proportional to the deviation of the current state of the system from the zero solution; therefore, the stochastic version of (4) is in the form
$$dX(t) = \left( pX(t-\tau) e^{-aX(t-\tau)} - \delta X(t) \right) dt + \lambda X(t) dB(t).$$

The parameter $\lambda$ represents the intensity of the noise. Necessary and sufficient conditions of mean-square stability of the trivial equilibrium of (5) are studied using a new approach, which provides better areas of stability with some numerical simulations that strengthen the theoretical findings.

The plan of the paper is as follows: In Sect. 2, some important preliminaries and notations are introduced. In Sect. 3, we introduce our new concept by proving the mean-square asymptotic stability of the zero solution of (1), (3) accordingly (5). A detailed numerical example is introduced in Sect. 4, with stability regions and some numerical simulations and interpretation. Conclusions and future works are presented to close the paper in Sect. 5.

2 Preliminaries

Notations: [13, 34–36]

1. Define $S_h = \{x \in \mathbb{R}^n, \|x\| < h\}$ and $\mathcal{K}$ as the family of all continuous nondecreasing functions $\nu : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that $\nu(x) > 0 \forall x > 0$ and $\nu(0) = 0$.

2. $L^2_{\mathcal{F}_t}(\Omega; \mathbb{R})$ is the family of $\mathbb{R}$-valued $\mathcal{F}_t$-measurable random variables $\xi$ such that
$$E[\xi]^2 < \infty.$$ 

In general, a stochastic process $\{X(t), t \geq 0\}$ defined on the probability space $(\Omega, \mathcal{F}, \mathbb{P})$ is called a second-order stochastic process if $X(t)$ is a $2$-r.v $\forall t \geq 0$. Then,
$$E[X^2(t)] < \infty.$$ 

3. $L^2([a, b]; \mathbb{R})$ is the family of $\mathbb{R}$-valued $\mathcal{F}_t$-adapted square-integrable processes $\{X(t)\}_{a \leq t \leq b}$ such that
$$\int_a^b |X(t)|^2 dt < \infty, \text{ a.s.}$$

4. Let $\mathcal{M}^2([0, T], \mathbb{R})$ denote the family of processes in $L^2([a, b]; \mathbb{R})$ such that
$$E\left( \int_a^b |X(t)|^2 dt \right) < \infty.$$ 

Definition 1 [37] The Lyapunov function $V(t, X) : [0, \infty) \times S_h \rightarrow \mathbb{R}_+$ is positive definite if $V(t, 0) \equiv 0$ and $V(t, X) \geq v(X)$ where $v(X) : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ is a nondecreasing positive-definite function, i.e., $v(0) \equiv 0$ and $v(X) > 0$ for $X \neq 0$.

Definition 2 [38] The standard one-dimensional Brownian motion $B(t)$ is a real-valued continuous $\{\mathcal{F}_t\}$-adapted process defined on the complete probability space $(\Omega, \mathcal{F}, \mathbb{F}^B, \mathbb{P})$ with the properties

1. $B(0) = 0$ a.s.
2. The increment $B(t) - B(s) \sim N(0, t - s)$ for $0 \leq s < t < \infty$ and independent of $\mathcal{F}_s$.

Lemma 1 [34] Assume $\{X(t), t \geq 0\}$ is adapted stochastic process defined on the underlying complete filtered probability space $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \geq 0}, \mathbb{P})$ where $\mathcal{F}_t$ is the filtration generated by the process $X(t)$. Let $X(t) \in \mathcal{M}^2([a, b], \mathbb{R})$, then for $a \leq t_0 \leq t_1 < b$
$$E\left( \int_a^b X(s) dB_s \right) = 0.$$ 

Proof The stochastic process $X(t)$ is defined on $\mathcal{M}_0([a, b], \mathbb{R})$, i.e., it is a simple or step process if there exists the partition $a = t_0 < t_1 < \cdots < t_k = b$
and $\psi_i, 1 \leq i \leq k$ are bounded random variables such that $\psi_i$ are $\mathcal{F}_{t_i}$-measurable, then

$$X(t) = \psi_0 I_{[0,t_1]} + \sum_{i=1}^{k-1} \psi_i I_{[t_i,t_{i+1}]}.$$  

Clearly, $\mathcal{M}_0([a, b], \mathbb{R}) \subset \mathcal{M}^2([a, b], \mathbb{R})$. Then

$$\mathbb{E} \int_a^b X(s) dB_s = \sum_{i=1}^k \mathbb{E} \left[ \psi_i (B_{t_{i+1}} - B_{t_i}) \right] = \sum_{i=1}^k \mathbb{E} [\psi_i] \mathbb{E} \left[ B_{t_{i+1}} - B_{t_i} \right] = 0,$$

where $B_{t_{i+1}} - B_{t_i}$ is independent of $\mathcal{F}_{t_i}$.

**Definition 3** [39, 40] The zero solution of (1) is

1. Mean square stable if for each $\varepsilon > 0$, $\exists \delta > 0$ and $\|\phi\|^2 < \delta$ such that

$$\mathbb{E} \|X(t, t_0, \phi)\|^2 < \varepsilon.$$

2. Asymptotically mean square stable if it is mean square stable and

$$\lim_{t \to \infty} \mathbb{E} \|X(t, t_0, \phi)\|^2 = 0.$$

**3 Main result**

Our main finding is to investigate the stability of the zero solution of the general Eqs. (1), (3) in the sense of the mean square using a suitable quadratic form of Lyapunov functional. Next theorems provide the mean-square stability under a constructed deterministic function in the neighborhood of the origin. This way is more effective and provides different stability conditions. Based on this result, the extinction of the Nicholson’s blowflies species will be studied.

**Theorem 1** The trivial solution of (1) is mean square asymptotically stable if there exists a negative definite functional $\theta(t, X_t)$ in the neighborhood of the origin such that

$$\theta(t, X_t) = 2X^T(t) Q \mu(t, X_t) + \sigma^T(t, X_t) Q_d \sigma(t, X_t).$$  

**Proof** Define the Lyapunov functional in the quadratic form

$$\mathcal{V}(t, X_t) = X^T(t) Q X(t),$$  

where $Q$ is a positive-definite symmetric matrix, $T$ is the transposition and define $Q_d$ as a diagonal matrix that has the same element of the diagonal of $Q$.

$$d\mathcal{V}(t, X_t, X(t)) = \mathcal{V}(t, X_t, X(t) + dX(t)) - \mathcal{V}(t, X_t, X(t))$$

$$= \left[ X^T(t) + \mu^T(t, X_t) \right] d\mathcal{V}(t, X_t) + \left[ \sigma(t, X_t) dB(t) \right]^T Q \left[ \sigma(t, X_t) dB(t) \right]$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right]$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right]$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right].$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right]$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right]$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right].$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right].$$

$$= \left[ X^T(t) Q X(t) + X^T(t) Q \mu(t, X_t) dX(t) + X^T(t) Q \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t) \right].$$

Since $B(t) \sim N(0, t)$ and

$$\begin{cases} 
\mathbb{E} [dB(t)] = 0, \\
\mathbb{E} [dB^2(t)] = dt, \\
\mathbb{E} [dt^2] = 0, \\
\mathbb{E} [dB(t) dt] = 0,
\end{cases}$$

hold, then taking the expectation implies

$$\mathbb{E} [d\mathcal{V}(t, X_t, X(t))] = 2X^T(t) Q \mu(t, X_t) dt + \sigma(t, X_t) dB(t) + Q \sigma(t, X_t) dB(t)$$

$$= \left[ 2X^T(t) Q \mu(t, X_t) + \sigma^T(t, X_t) Q_d \sigma(t, X_t) \right] dt$$

$$= \theta(t, X_t) dt,$$

where $\theta(t, X_t) = 2X^T(t) Q \mu(t, X_t) + \sigma^T(t, X_t) Q_d \sigma(t, X_t)$, and it is a negative-definite deterministic functional defined on the neighborhood of the origin; therefore, for $M > 0$

$$\mathbb{E} [\theta(t, X_t)] = \theta(t, X_t),$$

$$\theta(t, X_t) \leq -M \mathbb{E} [\mathcal{V}(t, X_t, X(t))].$$

$$\theta(t, X_t) = \frac{\mathbb{E} [d\mathcal{V}(t, X_t, X(t))]}{dt} \leq -M \mathbb{E} [\mathcal{V}(t, X_t, X(t))].$$

Then

$$\mathbb{E} [\mathcal{V}(t, X_t, X(t))] \leq e^{-Mt},$$

where $Q$ is a positive-definite symmetric matrix, $T$ is the transposition and define $Q_d$ as a diagonal matrix that has the same element of the diagonal of $Q$.
i.e.,
\[
\lim_{t \to \infty} \mathbb{E} [\mathcal{V}(t, X_t, X(t))] = \lim_{t \to \infty} \mathbb{E} \left[ X^2(t) \right]
\]
\[
= \lim_{t \to \infty} \mathbb{E} \left[ X^T(t) Q X(t) \right] = 0.
\]

Therefore, the zero solution of (1) is asymptotically mean square stable. □

**Theorem 2** The trivial solution of (3) is mean-square asymptotically stable if there exists a negative-definite functional \( \theta(t, X_t) \) in the neighborhood of the origin such that
\[
\theta(t, X_t) = \left[ X^T(t) + X_t^T \right] A^T Q X(t) + X^T(t) Q A [X(t) + X(t - \tau)]
\]
\[
+ X^T(t) Q A [X(t) + X(t - \tau)] + X^T(t) C^T Q_d C X(t).
\]

**Proof** Consider the linear part of (3) as a special case by letting
\[
\mu(t, X_t) = A (X(t) + X(t - \tau)),
\]
\[
\sigma(t, X_t) = C X(t).
\]

Then, the linear system becomes
\[
dX(t) = A (X(t) + X(t - \tau)) dt + C X(t) dB(t),
\]
where \( A, C \) are constant \( m \times m \) matrices. Now, assume the Lyapunov functional (7)
\[
d\mathcal{V}(t, X_t, X(t)) = \mathcal{V}(t, X_t, X(t) + dX(t)) - \mathcal{V}(t, X_t, X(t))
\]
\[
= \left[ X^T(t) + (A (X(t) + X(t - \tau)))^T dt + (C X(t) dB(t))^T \right] Q
\]
\[
\left[ X(t) + (A (X(t) + X(t - \tau))) dt + C X(t) dB(t) \right]
\]
\[
- X^T(t) Q X(t)
\]
\[
= \left[ X^T(t) + X^T(t) A^T dt + X^T(t - \tau) A^T dt 
\right.
\]
\[
\left. + (C X(t) dB(t))^T \right] Q
\]
\[
\left[ X(t) + A X(t) dt + A X(t - \tau) dt + C X(t) dB(t) \right]
\]
\[
- X^T(t) Q X(t).
\]
Using properties (8), then
\[
\mathbb{E} [d\mathcal{V}(t, X_t, X(t))] = X^T(t) Q A X(t) dt 
\]
\[
+ X^T(t - \tau) A^T dt Q X(t) + X^T(t) Q A X(t - \tau) dt 
\]
\[
+ X^T(t) A^T dt Q X(t)
\]
\[
+ (C X(t) dB(t))^T Q C X(t) dB(t)
\]
\[
= \left\{ \left[ X^T(t) + X^T(t - \tau) \right] A^T Q X(t) 
\right.
\]
\[
\left. + X^T(t) Q A [X(t) + X(t - \tau)] 
\right.
\]
\[
\left. + X^T(t) C^T Q_d C X(t) \right\} dt.
\]

Following the proof of previous theorem, assume
\[
\theta(t, X_t) = \left[ X^T(t) + X_t^T \right] A^T Q X(t)
\]
\[
+ X^T(t) Q A [X(t) + X_t] + X^T(t) C^T Q_d C X(t).
\]
\[
\mathbb{E} [d\mathcal{V}(t, X_t, X(t))] \leq 0 \text{ if } \theta(t, X_t) \text{ is negative-definite deterministic functional defined on the neighborhood of the origin, i.e., for } M > 0,
\]
\[
\mathbb{E} [\theta(t, X_t)] = \mathbb{E} [d\mathcal{V}(t, X_t, X(t))] 
\]
\[
\leq -M \mathbb{E} [\mathcal{V}(t, X_t, X(t))].
\]

Then,
\[
\mathbb{E} [\mathcal{V}(t, X_t, X(t))] \leq e^{-Mt},
\]
i.e.,
\[
\lim_{t \to \infty} \mathbb{E} [\mathcal{V}(t, X_t, X(t))] = \lim_{t \to \infty} \mathbb{E} \left[ X^2(t) \right]
\]
\[
= \lim_{t \to \infty} \mathbb{E} \left[ X^T(t) Q X(t) \right] = 0.
\]

Therefore, the zero solution of (3) is asymptotically mean square stable. □

**Proposition 1** The trivial solution of (5) is mean square asymptotically stable if
\[
2p X^T Q X_t e^{-\alpha X_t} + \lambda^2 X^T Q_d X(t) \leq 2\delta X^T Q X(t)
\]

**Proof** The proof follows directly from Theorem 1, by considering the same quadratic Lyapunov functional (7) and using properties (8), and leads to the functional
\[
\theta(t, X_t) = 2 \left[ p X^T Q X_t e^{-\alpha X_t} 
\right.
\]
\[
\left. + \frac{\lambda^2}{2} X^T Q_d X(t) - \delta X^T Q X(t) \right]
\]
which should be negative definite in the neighborhood of the origin for asymptotic mean-square stability. □
Now, centering system (5) on the zero equilibrium solution $X_0 = 0$, linearizing using the notation $e^{-aX(t-\tau)} = 1 - aX(t-\tau) + \mathcal{O}(X(t-\tau))$, and neglect $\mathcal{O}(X(t-\tau))$ by $\lim_{\chi \to 0} \frac{\mathcal{O}(X)}{X} \to 0$, we get the corresponding process $Y(t)$

$$dY(t) = (pY(t - \tau) - \delta Y(t)) \, dt + \lambda Y(t) \, dB(t). \quad (11)$$

**Proposition 2** The trivial solution of (11) is mean-square asymptotically stable if

$$2X^T Q (p X_t - \delta X(t)) + \lambda^2 X^T Q dX(t) \leq 2 \left( \delta X^T(t) - p X_t^T \right) \delta X^T Q X$$

Proof The proof follows from Theorem 2, by considering the same-quadratic Lyapunov functional (7) and using properties (8), and leads to the functional

$$\theta(t, X_t) = 2 \left[ X^T Q (p X_t - \delta X(t)) + \frac{\lambda^2}{2} X^T Q dX(t) \right]$$

which should be negative definite in the neighborhood of the origin for asymptotic mean-square stability for the linear system (11) and consequently for (5). □

### 4 Example

Consider the linear system (11). By introducing the quadratic Lyapunov functional (7) for $Q = 1$ and $V(t, X_t) = V_1 + V_2$, where

$$V_1(t, X_t, X(t)) = X^T(t) X(t) = X^2(t).$$

Then, from Proposition 1, and system (11)

$$\mathbb{E} [dV_1(t, X_t, X(t))] = \mathbb{E} [2X(t) (p X(t - \tau) - \delta X(t)) + \lambda^2 X^2(t) \, dt]$$

$$\leq \mathbb{E} [p X^2(t) + p X^2(t - \tau) - 2\delta X^2(t) + \lambda^2 X^2(t)] \, dt$$

$$= (p - 2\delta + \lambda^2) \mathbb{E} [X^2(t)] \, dt + p \mathbb{E} [X^2(t - \tau)] \, dt.$$

For the negative definiteness, choose

$$V_2(t, X_t, X(t)) = p \int_{t-\tau}^t X^T(s) X(s) \, ds$$

$$ds = p \int_{t-\tau}^t X^2(s) \, ds.$$

Then

$$\mathbb{E} [dV(t, X_t, X(t))] \leq (2p - 2\delta + \lambda^2) \mathbb{E} [X^2(t)] \, dt.$$ According to Theorem 1, for $Q = 1$

$$\theta(t, X_t) = (2p - 2\delta + \lambda^2) \mathbb{E} [X^2(t)],$$

which is negative definite if

$$2p + \lambda^2 < 2\delta. \quad (12)$$

Condition (12) is a necessary condition for the mean-square stability of the zero solution of (5); it is a delay-independent condition. Accordingly, Fig. 1 shows the regions of mean-square stability.

By introducing different Lyapunov functionals, we get different stability conditions. Assume another Lyapunov functional in the form

$$V(t, X_t, X(t)) = \left( X(t) + p \int_{t-\tau}^t X(s) \, ds \right)^2$$

$$+ p |p - \delta| \int_{t-\tau}^t \int_0^s X^2(s) \, ds.$$

Then,

$$\mathbb{E} [dV(t, X_t, X(t))] = 2 \mathbb{E} \left[ \left( X(t) + p \int_{t-\tau}^t X(s) \, ds \right)^2 \right]$$

$$\leq p \mathbb{E} [p X^2(t) + p X^2(t - \tau) - 2\delta X^2(t) + \lambda^2 X^2(t)] \, dt$$

$$\leq (2p - 2\delta + \lambda^2) \mathbb{E} [X^2(t)] \, dt$$

$$+ 2p |p - \delta| \mathbb{E} [X^2(t) \int_{t-\tau}^t X(s) \, ds] \, dt$$

$$- p |p - \delta| \mathbb{E} \left[ \int_{t-\tau}^t X^2(s) \, ds \right] \, dt$$

$$\leq (2p - 2\delta + 2p|p - \delta| + \lambda^2) \mathbb{E} [X^2(t)] \, dt$$

$$+ p |p - \delta| \mathbb{E} \left( X^2(t) \int_{t-\tau}^t X^2(s) \, ds \right) \, dt$$

$$- p |p - \delta| \mathbb{E} \left[ \int_{t-\tau}^t X^2(s) \, ds \right] \, dt$$

Therefore,

$$\theta(t, X_t) = (2p - 2\delta + 2p|p - \delta| + \lambda^2) \mathbb{E} [X^2(t)]$$

is negative definite if

$$p \tau |p - \delta| + \frac{\lambda^2}{2} < \delta - p. \quad (14)$$
Mean-square stability of the zero equilibrium

This delay-dependent condition is necessary for the mean-square stability. Via this new condition, we can show the impact of delay on the stability regions with $\lambda$. Figure 2 shows the mean-square stability regions for different values of $\tau$ with fixed $\lambda = 1.8$. For fixed delay $\tau = 0.5$, mean-square stability regions are shown in Fig. 3 for different $\lambda$.

Regarding the numerical simulation, we use the Euler–Maruyama algorithm which has been shown by [1]. According to (2),

$$
\int_{t_n}^{t_{n+1}} \mu(s, x) ds \approx \mu(t_n, x_n) \Delta t, \quad \text{and} \quad \int_{t_n}^{t_{n+1}} \sigma(s, x) dB(s) \approx \sigma(t_n, x_n) \Delta B_n.
$$

Then, the scheme of EM has the form

$$
x_{n+1} = x_n + \mu(t_n, x_n) \Delta t + \sigma(t_n, x_n) \Delta B_n,
$$

where $x(t_n)$ is the data required in the scheme and $x(t_{n+1})$ is the resulting process at $t_{n+1}$. The EM scheme is strongly convergent with order 0.5. We choose the stepsize $\Delta t$ appropriately to avoid time and errors of computation. For the initial history function $\phi = 2.5 \cos(3s), \ s \in [-3.5, 0]$ and $\lambda = 1.8, \ \tau = 3.5$, Fig. 4a shows stable trajectories of the solution of (5), and conditions (12), (14) are satisfied. If these conditions are not met, then we get unstable trajectories of the zero solution as shown in Fig. 4b.

The impact of the intensity of the noise $\lambda$ and the delay in the production process $\tau$ on the behavior of the solution should be noted. Figure 5 shows three different behaviors of the solution according to condition (14). The impact of $\tau$ is shown in Fig. 6. For small values of $\lambda, \tau$ in the light of condition (14), we get asymptotic stable solutions.
4.1 Interpretation

By introducing a suitable Lyapunov functional in the previous example, we have obtained a delay-independent condition (12) that gives areas of mean-square stability of the zero solution in \((p, \delta)\) space of parameters for different values of \(\lambda\). By choosing small values of the intensity of the noise, we get better areas of stability as shown in Fig. 1. By introducing a different Lyapunov functional, we have obtained a delay-dependent condition (14), which is better as it considers the variation in the delay of the production process of eggs. We investigate the impact of \(\lambda\), and we arrive at the same result of better regions by decreasing \(\lambda\); this is shown in Fig. 3. Moreover, choosing small values of delay \(\tau\) helps in the extinction of the Nicholson’s blowflies, and better regions of stability are obtained for small \(\tau\). Some computer simulations are carried out to support the result. Choosing the point \((0.1, 3.0)\) which is in the stability area, performing the numerical simulation of the solution with \(X = 0\), we get 25 blue stable trajectories in Fig. 4a. Figure 4b shows 25 red unstable trajectories of the solution by choosing the coordinate \((3.0, 1.0)\) which is out of stability region, and accordingly condition (14) is not satisfied. Figure 5 shows the impact of the environmental noise on the behavior.
5 Conclusion and further directions

Our work has led us to conclude the necessary and sufficient conditions for stability of the zero solution in mean square under the influence of white noise term. This new approach can provide different stability (delay-dependent) conditions. This work can be extended to many applications in many disciplines; for instance, neural networks, infectious diseases and the generalized system of animal population such as the general equation of the Nicholson’s blowflies model, which is known as the neoclassical growth model

\[ dX(t) = \left( pX(t - h) e^{-\alpha X(t-h)} - \delta X(t) \right) dt + \lambda X(t) dB(t). \]

Of course, our concept of stability has some limitations,

1. It is not yet known whether this approach can be applied to systems involving the general fractional Brownian motion \( B_H(t), 0 < H < 1 \), like the applications studied by [41,42], for instance.

2. A challenging point to us is to study the mean-square stability using our approach in case of systems involving distributed delays.

Acknowledgements Not applicable.

Author contributions Not applicable.

Funding This work was supported by the Mathematics Department—Mansoura University of Egypt.

Availability of data and materials The data sets generated and/or analyzed during the current study are available from the corresponding author on reasonable request.

Declarations

Conflict of interest The author declares that there is no conflict of interests regarding the publication of this article.

References

1. Maruyama, G.: Continuous markov processes and stochastic equations. Rendiconti del Circolo Matematico di Palermo 4(1), 48 (1955)
2. Cao, W., Liu, M., Fan, Z.: Ms-stability of the euler-maruyama method for stochastic differential delay equations. Appl. Math. Comput. 159(1), 127–135 (2004)
3. Gopalsamy, K.: Stability and Oscillations in Delay Differential Equations of Population Dynamics, 74th edn. Springer, Berlin (2013)
4. Győri, I.: Delay differential and integro-differential equations in biological compartment models. Syst. Sci. 8(2–3), 167–187 (1982)

5. Kuang, Y.: Delay Differential Equations: With Applications in Population Dynamics. Academic Press, Cambridge (1993)

6. Brännström, Å.: Modelling Animal Populations: Tools and Techniques. Doctoral thesis. Umeå University, Faculty of Science and Technology, Mathematics and Mathematical Statistics (2004)

7. Rodney David, D.: Ordinary and Delay Differential Equations, 20th edn. Springer, Berlin (2012)

8. Erneux, T.:Applied Delay Differential Equations, 3rd edn. Springer, Berlin (2009)

9. Gopalsamy, K., Zhang, B.G.: On delay differential equations with impulses. J. Math. Anal. Appl. 139(1), 110–122 (1989)

10. Hale, J.K., Verduyn Lunel, S.M., Verduyn, L.S., Lunel, S.M.: Introduction to Functional Differential Equations, vol. 99. Springer, Berlin (1993)

11. Makay, G.: On the asymptotic stability of the solutions of functional differential equations with infinite delay. J. Differ. Equ. 108(1), 139–151 (1994)

12. Taylor, S.R.: Probabilistic Properties of Delay Differential Equations. arXiv preprint arXiv:1909.02544 (2019)

13. El-Metwally, H., Sohaly, M.A., Elbaz, I.M.: Stochastic global exponential stability of disease-free equilibrium of hiv/aids model. Eur. Phys. J. Plus 135(10), 1–14 (2020)

14. Mao, X.: Numerical solutions of stochastic functional differential equations. LMS J. Comput. Math. 6, 141–161 (2003)

15. Ma, L., Ning, X., Huo, X., Zhao, X.: Adaptive finite-time output-feedback control design for switched pure-feedback nonlinear systems with average dwell time. Nonlinear Anal. Hybrid Syst. 37, 100908 (2020)

16. Cai, J., Rui, Y., Wang, B., Mei, C., Shen, L.: Decentralized event-triggered control for interconnected systems with unknown disturbances. J. Franklin Inst. 357(3), 1494–1515 (2020)

17. Shaikhet, L.: Lyapunov Functionals and Stability of Stochastic Functional Differential Equations. Springer, Berlin (2013)

18. Wang, W., Shi, C., Chen, W.: Stochastic nicholson-type delay differential system. Int. J. Control 94, 1–8 (2019a)

19. Wang, W., Wang, L., Chen, W.: Stochastic Nicholson’s blowflies delayed differential equations. Appl. Math. Lett. 87, 20–26 (2019b)

20. Blythe, S., Mao, X., Liao, X.: Stability of stochastic delay neural networks. J. Franklin Inst. 338(4), 481–495 (2001)

21. Park, J.H., Kwon, O.M.: Analysis on global stability of stochastic neural networks of neutral type. Mod. Phys. Lett. B 22(32), 3159–3170 (2008)

22. Zhou, L., Guanda, H.: Almost sure exponential stability of neutral stochastic delayed cellular neural networks. J. Control Theory Appl. 6(2), 195–200 (2008)

23. Huang, C., Yang, X., Cao, J.: Stability analysis of nicholson’s blowflies equation with two different delays. Math. Comput. Simul. 171, 201–26 (2020)

24. Bradul, N., Shaikhet, L.: Stability of the positive point of equilibrium of nicholson’s blowflies equation with stochastic perturbations: numerical analysis. Discrete Dyn. Nature Soc. 2007, 1–26 (2007)

25. Shaikhet, L.: Stability of equilibriums of stochastically perturbed delay differential neoclassical growth model. Discrete Contin. Dyn. Syst., Ser. B 22(4), 1565–1573 (2017)

26. Berezansky, L., Idels, L., Troib, L.: Global dynamics of nicholson-type delay systems with applications. Nonlinear Anal. Real World Appl. 12(1), 436–445 (2011)

27. Gikhman, I.I., Skorokhod, A.V.: The Theory of Stochastic Processes II. Springer, Berlin (2004)

28. Kolmanovskii, V., Myshkis, A.: Introduction to the Theory and Applications of Functional Differential Equations, 463rd edn. Springer, Berlin (2013)

29. Gurney, W.S.C., Blythe, S.P., Nisbet, R.M.: Nicholson’s blowflies revisited. Nature 287(5777), 17–21 (1980)

30. Shu, H., Wang, L., Wu, J.: Global dynamics of nicholson’s blowflies equation revisited, onset and termination of non-linear oscillations. J. Differ. Equ. 255(9), 2565–2586 (2013)

31. Berezansky, L., Braverman, E., Idels, L.: Nicholson’s blowflies differential equations revisited: main results and open problems. Appl. Math. Model. 34(6), 1405–1417 (2010)

32. Wang, W., Wang, L., Chen, W.: Existence and exponential stability of positive almost periodic solution for Nicholson-type delay systems. Nonlinear Anal. Real World Appl. 12(4), 1938–1949 (2011)

33. Van Hien, L.: Global asymptotic behaviour of positive solutions to a non-autonomous Nicholson’s blowflies model with delays. J. Biol. Dyn. 8(1), 135–144 (2014)

34. Mao, X.: Stochastic Differential Equations and Applications. Elsevier, Amsterdam (2007)

35. Sohaly, M.A., Yassen, M.T., Elbaz, I.M.: Stochastic consistency and stochastic stability in mean square sense for cauchy advection problem. J. Differ. Equ. Appl. 24(1), 59–67 (2018)

36. Villafuerte, L., Braumann, C.A., Cortés, J.C., Jódar, L.: Random differential operational calculus: theory and applications. Comput. Math. Appl. 59(1), 115–125 (2010)

37. Quying, L.: Stability of sirs system with random perturbations. Physica A 388(18), 3677–3686 (2009)

38. Evans, L.C.: An Introduction to Stochastic Differential Equations, vol. 82. American Mathematical Society, Providence (2012)

39. Mao, X.: Exponential Stability of Stochastic Differential Equations. Marcel Dekker, New York (1994)

40. Mohammed, S.-E.A.: Stochastic Functional Differential Equations, Vol. 99. Pitman Advanced Publishing Program, Boston, London, Melbourne (1984)

41. Biagini, F., Hu, Y., Øksendal, B., Zhang, T.: Stochastic Calculus for Fractional Brownian Motion and Applications. Springer, Berlin (2008)

42. Mishura, I.S., Misura, J.S., Mishura, Y., Misura, I.S., Misura, Û.S.: Stochastic Calculus for Fractional Brownian Motion and Related Processes, vol. 1929. Springer, Berlin (2008)

Publisher’s Note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.