Dynamics of a stage-structured single population model with state-dependent delay

Yan Wang¹, Xianning Liu¹* and Yangjiang Wei²

¹Correspondence: liuxn@swu.edu.cn
²Key Laboratory of Eco-environments in Three Gorges Reservoir Region (Ministry of Education), School of Mathematics and Statistics, Southwest University, Chongqing, China
Full list of author information is available at the end of the article

Abstract

In this paper, a novel stage-structured single population model with state-dependent maturity delay is formulated and analyzed. The delay is related to the size of population and taken as a non-decreasing differentiable bounded function. The model is quite different from previous state-dependent delay models in the sense that a correction term, \(1 - \tau'(z(t))z(t)\), is included in the maturity rate. Firstly, positivity and boundedness of solutions are proved without additional conditions. Secondly, existence of all equilibria and uniqueness of a positive equilibrium are discussed. Thirdly, local stabilities of the equilibria are obtained. Finally, permanence of the system is analyzed, and explicit bounds for the eventual behaviors of the immature and mature populations are established.

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1 Introduction

In a natural ecosystem, the individual members of the population have a life history that takes them through two or more stages, especially, with regard to mammalian populations, which usually exhibit two distinct stages: immature and mature stages [1–4]. What is more, the delay arises frequently as the maturation time from birth to adulthood under the background of population dynamics, and there are different types of the maturity time delay, such as discrete, distributed, stochastic, time-dependent and state-dependent delays, etc. [4–11].

The authors of [12] considered and analyzed a growth model of blowflies with constant maturity time delay. According to the data of blowflies growth experiments in [13], they presented some numerical simulations to justify the theoretical analysis. Thus, it is more practical to consider time delays and stage structures in the study of a population model. In 1990, the authors of [14] developed and analyzed the following stage structure model of population growth with a constant maturity time delay:

\[
\begin{align*}
\frac{dx(t)}{dt} &= ax(t) - bx(t) - cy(t - \tau)e^{-\gamma t}, \\
\frac{dy(t)}{dt} &= ay(t) e^{-\gamma t} - \beta y^2(t),
\end{align*}
\]

(1.1)
where \( x(t) \) and \( y(t) \) represent the immature and mature population densities, respectively; the time delay \( \tau \) represents the time from birth to maturity; parameters \( \alpha \) and \( \gamma \) represent the birth and death rates of the immature population, respectively; \( \beta \) represents the mature death and overcrowding rate.

Before World War II, people discovered that juvenile seals needed five years to mature, small whales needed seven to ten years, and big whales needed 12 to 15 years. After the war, owing to the introduction of factory ships and capture of a large number of whales, the number of the krill available for the seals and the remaining whales had a significant increase. An interesting phenomenon was then observed that the maturity time of seals and whales has shortened. Seals needed three to four years and small whales only took five years. Additionally, there was a significant reduction of maturity time for large whales [15]. Therefore, for Antarctic whale and seal populations, their maturity time is a function of the number of krill. Furthermore, it is quite natural to suppose that the number of krill available in a closed environment is a function of the total population.

This suggests that due to the environmental complexity, however, the time delay may be constantly adjusted as the state changes, that is, the time delay is state-dependent and the constant time delay is no longer reasonable (see [16–19]). In view of this, on the basis of model (1.1), the authors of [5] changed the constant delay \( \tau \) into a state-dependent delay \( \tau(z(t)) \) to formulate the following state-dependent delay model:

\[
\begin{align*}
\frac{dx(t)}{dt} &= \alpha y(t) - \gamma x(t) - \alpha y(t - \tau(z(t)))e^{-\gamma \tau(z(t))}, \\
\frac{dy(t)}{dt} &= \alpha y(t - \tau(z(t)))e^{-\gamma \tau(z(t))} - \beta y^2(t),
\end{align*}
\] (1.2)

where the state-dependent time delay \( \tau(z(t)) \) is taken to be an increasing differentiable bounded function of the total population \( z(t) = x(t) + y(t) \).

Since then, more and more state-dependent delay differential equations were applied to describe the dynamical behaviors of stage-structured models originated from population growth [19–25].

In 2004, based on model (1.2), the authors of [26] studied the following stage-dependent population model with a state-dependent time delay:

\[
\begin{align*}
\frac{dx(t)}{dt} &= R(y(t)) - \gamma x(t) - R(y(t - \tau(z(t))))e^{-\gamma \tau(z(t))}, \\
\frac{dy(t)}{dt} &= R(y(t - \tau(z(t))))e^{-\gamma \tau(z(t))} - \beta y(t),
\end{align*}
\] (1.3)

where the immature birth rate \( R(y(t)) \) is taken as a general function of the present mature population and the death rate for the mature one is a constant.

Subsequently, [27] and [28] investigated a two cooperative mature species system with a state-dependent delay and a cooperative model composed of two species with a stage structure and state-dependent maturity delays, respectively.

However, it is obvious that those state-dependent delay equations are all directly replacing constant delays in classical models of the form

\[
\frac{dX(t)}{dt} = F(X(t - \tau), X(t)), \quad \tau > 0,
\]
by state-dependent delays, hence one obtains the following equations:

\[
\frac{dX(t)}{dt} = F(X(t - \tau(X(t))), X(t)).
\]

Is this approach appropriate and in line with population modeling? Since if the maturity delay is not constant but state-dependent, then changes in the number of mature individuals do not only depend on reproduction and death, but also on the changing definition of maturity, that is, changes of \( \tau(X(t)) \). As we shall show later, the correct extension of model (1.2) in terms of a state-dependent mature delay involves a correction term, \( 1 - \tau'(z(t))\dot{z}(t) \), which takes into account the aforesaid changes.

Moreover, Macdonald [29] indicated that one must begin with an age structured model to incorporate maturation data in a model, which is indispensably formulated in terms of partial differential equations. This way it is guaranteed that the model can properly be substituted by that formulated according to a functional differential equation.

Our paper is organized as follows. In Sect. 2, we formulate a novel species stage-structured model with a state-dependent maturation delay. In Sect. 3, we discuss the positivity and boundedness of solutions, as well as the existence of all equilibria and uniqueness of a positive equilibrium. In Sect. 4, we analyze the linear stability of equilibria. In Sect. 5, we discuss the permanence of the system. Finally, in Sect. 6, we briefly discuss and summarize our results.

2 Model formulation and hypotheses

2.1 Model formulation

Motivated by [8, 29, 30], we begin with a two-stage population model (immature and mature stages). In order to distinguish immature individuals, \( x(t) \), from mature ones, \( y(t) \), we introduce a threshold age \( \tau(z(t)) \), which is the maturation time for an immature individual that matures at time \( t \) depending on the total population \( z(t) = x(t) + y(t) \). Let \( \rho(t, a) \) be the density of population of age \( a \) at time \( t \). Then the number of immature individuals, \( x(t) \), and mature individuals, \( y(t) \), respectively, are given by

\[
x(t) = \int_{0}^{\tau(z(t))} \rho(t, a) \, da \quad \text{and} \quad y(t) = \int_{\tau(z(t))}^{\infty} \rho(t, a) \, da.
\]

The evolution of population is represented by the following age structure partial differential equations [31, 32]:

\[
\begin{align*}
\frac{\partial \rho(t, a)}{\partial t} + \frac{\partial \rho(t, a)}{\partial a} &= -\gamma \rho(t, a), \quad \text{if } a \leq \tau(z(t)), \\
\frac{\partial \rho(t, a)}{\partial t} + \frac{\partial \rho(t, a)}{\partial a} &= -\beta y(t) \rho(t, a), \quad \text{if } a > \tau(z(t)),
\end{align*}
\]

(2.1)

where each individual from \( x(t) \) dies at a constant rate \( \gamma \) and that from \( y(t) \) at a non-constant rate \( \beta y(t) \).

Taking the derivatives of \( x(t) \) and \( y(t) \), respectively, and combining with (2.1), we get

\[
\begin{align*}
\frac{dx(t)}{dt} &= \rho(t, 0) - \gamma x(t) - \left[ 1 - \tau'(z(t))\dot{z}(t) \right] \rho(t, \tau(z(t))), \\
\frac{dy(t)}{dt} &= \left[ 1 - \tau'(z(t))\dot{z}(t) \right] \rho(t, \tau(z(t))) - \rho(t, \infty) - \beta y(t).
\end{align*}
\]
It is necessary to note that a prime refers to differentiation with respect to $z$, and a dot indicates differentiation with respect to time $t$, namely, $\dot{z}(t) = d\tau(z(t))/dt = \tau'(z(t))\dot{z}(t)$.

Because any individual cannot live forever, $\rho(t, \infty)$ is regarded as zero. We assume that the immature population produces at a constant rate $\alpha$, so the term $\rho(t, 0) = \alpha y(t)$ represents the number of immature individuals born at time $t$. Therefore, for $t \geq \tilde{t} = \max\{\tau(z(t))\}$, we obtain

$$\rho(t, \tau(z(t))) = \rho(t - \tau(z(t)), 0) = \alpha y(t - \tau(z(t))) e^{-\gamma\tau(z(t))}.$$

As a result, we have the following stage-structured population model with a state-dependent delay:

$$\begin{align*}
\frac{dx(t)}{dt} &= \alpha y(t) - \gamma x(t) - \alpha [1 - \tau'(z(t))\dot{z}(t)] y(t - \tau(z(t))) e^{-\gamma\tau(z(t))}, \\
\frac{dy(t)}{dt} &= \alpha [1 - \tau'(z(t))\dot{z}(t)] y(t - \tau(z(t))) e^{-\gamma\tau(z(t))} - \beta y^2(t).
\end{align*}$$

(2.2)

Model (2.2) is clearly different from the previous state-dependent delay equations in the sense that it includes the correction term $1 - \tau'(z(t))\dot{z}(t)$ in the maturity rate.

As in [5, 26, 28, 30], to have the possibility of mature individuals becoming immature only by birth, one needs some conditions ensuring that $t - \tau(z(t))$ is a strictly increasing function of $t$. As a matter of fact, from a biological point of view, it is natural that $t - \tau(z(t))$ is a strictly increasing function of $t$. Assume that $r(\xi)$ is the developmental proportion at time $\xi$. Then, when an immature individual moves to the mature state from $t - \tau(z(t))$ to $t$, the cumulative rate of development $r$ should be equal to one, namely,

$$1 = \int_{t-\tau(z(t))}^{t} r(\xi) \, d\xi.$$

Taking the derivative with respect to $t$, we obtain

$$1 - \tau'(z(t))\dot{z}(t) = \frac{r(t)}{r(t - \tau(z(t)))} > 0.$$

It implies that $t - \tau(z(t))$ is a strictly increasing function of $t$ and the maturity time delay $\tau(z(t))$ doesn’t change arbitrarily over time.

### 2.2 Model hypotheses

The hypotheses for model (2.2) are as follows:

(A1) Parameters $\alpha, \gamma, \beta$ are all positive constants;

(A2) The state-dependent maturity time delay $\tau(z)$ is an increasing differentiable bounded function of the total population $z = x + y$, where $\tau'(z) \geq 0$, $\tau''(z) \leq 0$, and $0 < \tau_m \leq \tau(z) \leq \tau_M$ with $\tau(0) = \tau_m$ and $\tau(\infty) = \tau_M$.

Throughout this paper, we will do qualitative analysis for system (2.2). The initial conditions for system (2.2) are

$$\begin{align*}
y(s) &= \Phi(s) \geq 0, \\
x(s) &= \Psi(s) \geq 0 \quad \text{for all } s \in [-\tau_M, 0],
\end{align*}$$
with
\[
\Psi(0) = \int_{-\tau(z(0))}^{0} \alpha \Phi(s)e^{\gamma s} \, ds,
\]
which denotes the size of the immature population surviving to time \( t = 0 \), where \( \tau(z(0)) \) is the maturation time at \( t = 0 \), namely,
\[
\tau(z(0)) = \tau\left(\Phi(0) + \int_{-\tau(z(0))}^{0} \alpha \Phi(s)e^{\gamma s} \, ds\right).
\]

3 Preliminary results

In this section, we will discuss the positivity and boundedness of solutions and the existence of all equilibria and uniqueness of a positive equilibrium.

3.1 Positivity and boundedness

**Theorem 3.1** Let \( \Phi(t) \geq 0, \Psi(t) \geq 0 \) for \( -\tau_M \leq t \leq 0 \) and \( \Phi(0) > 0 \). Then \( y(t) > 0 \) and \( x(t) > 0 \) for all \( t \geq 0 \).

**Proof** Since \( 1 - \tau'(z(t)) \dot{z}(t) > 0 \), for \( -\tau_M \leq t \leq 0 \), we obtain
\[
\dot{y}(t) \geq -\beta y(t)^2.
\]
By comparison, \( y(t) \geq y_1(t) \) where \( y_1(t) \) is the solution of \( \dot{y}_1(t) = -\beta y_1(t)^2 \), \( y_1(0) = \Phi(0) > 0 \). Therefore, \( y(t) > 0 \) for all \( -\tau_M \leq t \leq 0 \). We can continue the argument to all positive times, so \( y(t) > 0 \) for all \( t \geq 0 \). Now let us prove the positivity of \( x(t) \). Integrating the first equation of (2.2), we get
\[
x(t) = e^{-\gamma t} \left( \Psi(0) + \int_{0}^{t} \alpha y(s)e^{\gamma s} \, ds - \int_{-\tau(z(0))}^{t-\tau(z(t))} \alpha y(s)e^{\gamma s} \, ds \right)
\]
\[
= e^{-\gamma t} \left( \int_{-\tau(z(0))}^{0} \alpha y(s)e^{\gamma s} \, ds + \int_{0}^{t} \alpha y(s)e^{\gamma s} \, ds - \int_{-\tau(z(0))}^{t-\tau(z(t))} \alpha y(s)e^{\gamma s} \, ds \right)
\]
\[
= \int_{t-\tau(z(t))}^{t} \alpha y(s)e^{\gamma s} \, ds.
\]
By the positivity of \( y(t) \) and \( z(z(t)) \), we have \( x(t) > 0 \). The proof is complete. \( \square \)

**Remark 3.1** In Theorem 3.5 of [5] and Theorem 2.6 of [28], one needs the following stringent conditions for \( \tau'(z(t)) \) to ensure the positivity of \( x(t) \):

(i) \( \tau'(z) < 4\beta/\alpha^2 \);

(ii) \( \tau'(z) > 0 \) is small enough so that the inequality

\[
K_m \int_{t-\tau_m}^{t} e^{\gamma s} \, ds > \Delta \int_{t-\tau}^{t-\tau_m} \frac{\alpha^2 \tau'(z)}{4\beta - \alpha^2 \tau'(z)} e^{\gamma s} \, ds,
\]
holds for all values of \( t \), where \( K_m > 0 \) and \( y(t) > K_m \) for all \( t \geq 0 \). As a matter of fact, according to Theorem 3.1, we see that neither of these conditions is necessary.
\textbf{Theorem 3.2} Let $\Phi(t) \geq 0$, $\Psi(t) \geq 0$ for $-\tau_M \leq t \leq 0$ and $\Phi(0) > 0$, then the solution $(x(t), y(t))$ of system (2.2) is uniformly ultimately bounded for $t \geq 0$.

\textit{Proof} Define a Lyapunov function as follows:

$$V(t) = x(t) + y(t).$$

Calculating the time derivative of $V(t)$ along the solutions of system (2.2), we obtain

$$\dot{V}(t) = \dot{x}(t) + \dot{y}(t)$$

$$= \alpha y(t) - \gamma x(t) - \beta y^2(t)$$

$$= -\gamma V(t) + (\alpha + \gamma) y(t) - \beta y^2(t)$$

$$\leq -\gamma V(t) + M,$$

where $M > 0$ is the maximum of the quadratic function $(\alpha + \gamma)y(t) - \beta y^2(t)$. Therefore, $\limsup_{t \to \infty} V(t) \leq M/\gamma$ and the solution of system (2.2) is uniformly ultimately bounded. \hfill \Box

Obviously, it implies that $z(t)$ is bounded, which will be used in the proof of Theorem 5.1. In addition, $x(t) \leq M/\gamma$ and $y(t) \leq M/\gamma$, which is the ultimate upper bound for all the solutions. For the upper bound of each solution, we have the following results.

\textbf{Theorem 3.3} Let $\Phi(t) \geq 0$ for $-\tau_M \leq t \leq 0$ and $\Phi(0) > 0$. Then there exists a $\Delta = \Delta(\Phi) > 0$ such that $y(t) \leq \Delta$ for $t \geq 0$, where $\Delta = \max\{\sup_{-\tau_M \leq t \leq 0} \Phi(t), \alpha \beta^{-1} e^{-\gamma \tau_m}\}$.

\textit{Proof} Our proof is divided into three cases.

Case (a). Firstly, we start with the case where both $y(t)$ and $z(t)$ are eventually monotonic. If $z(t)$ is eventually decreasing, since $z(t) = x(t) + y(t)$, then $y(t)$ is bounded. Suppose that both $y(t)$ and $z(t)$ are eventually increasing, i.e., $\dot{y}(t) \geq 0$ and $\dot{z}(t) \geq 0$ for all $t > T$ for some $T \geq 0$. Then for $t > T + \tau_M$

$$\dot{y}(t) = \alpha \left[1 - \tau'\left(z(t)\right)\dot{z}(t)\right] y(t - \tau\left(z(t)\right)) e^{-\gamma \tau\left(z(t)\right)} - \beta y^2(t)$$

$$\leq \alpha \left[1 - \tau'\left(z(t)\right)\dot{z}(t)\right] y(t) e^{-\gamma \tau\left(z(t)\right)} - \beta y^2(t)$$

$$\leq \alpha y(t) e^{-\gamma \tau\left(z(t)\right)} - \beta y^2(t),$$

since $y(t - \tau(z(t))) \leq y(t)$. This implies that for $t > T$, we have

$$y(t) \leq \alpha \beta^{-1} \left[1 - \tau'(z(t))\dot{z}(t)\right] e^{-\gamma \tau(z(t))}$$

$$\leq \alpha \beta^{-1} e^{-\gamma \tau_m}.$$
If $t_m = s_m$, then by using similar methods as in the case (a), we see that $y(t) \leq \alpha \beta^{-1} e^{-\gamma t_m}$ for all $t < t_m$.

If $t_m < s_m$ and $y(t_m) \leq y(s_m)$, then $\dot{y}(s_m) > 0$ for all $t < t_m$. Otherwise, there exists a $t \in (t_m, s_m)$ such that $\dot{y}(t) = 0$, which contradicts the definition of $t_m$. Therefore, we obtain

$$
\dot{y}(s_m) = \alpha \left[ 1 - \tau' (z(s_m)) \right] y(s_m - \tau (z(s_m))) e^{-\gamma \tau(z(s_m))} - \beta y^2(s_m)
\leq \alpha y(s_m) e^{-\gamma \tau s_m} - \beta y^2(s_m).
$$

This implies that $y(t) \leq \alpha \beta^{-1} e^{-\gamma s_m}$.

If $t_m < s_m$ and $y(s_m) \leq y(t_m)$, then it is clear that $y(t)$ is bounded above.

Case (c). We discuss the case where one of $y(t)$ and $z(t)$ is oscillatory and the other is eventually monotonic. Without loss of generality, suppose that $y(t)$ is oscillatory and $z(t)$ is eventually increasing since the other cases can be worked out similarly. Therefore, there is a sequence $(t_n)_{n=1}^\infty$ such that $\dot{y}(t_n) = 0$, $y(t_n)$ is a local maximum of $y(t)$, and $y(t) \leq y(t_n)$ for all $0 < t < t_n$. For the same sequence $(t_n)_{n=1}^\infty$ presented above, there exists an $N > 0$ such that $z(t_n) \geq 0$ from the eventual monotonicity of $z(t)$, and thus $y(t)$ is bounded above.

Therefore, choosing $\Delta(\Phi) = \max \{ \sup_{-\tau M \leq t \leq 0} \Phi(t), \alpha \beta^{-1} e^{-\gamma s_m} \}$, the proof is complete. □

**Theorem 3.4** Let $\Phi(t) > 0$ for $-\tau M \leq t \leq 0$. Then there exists a $\Theta(\Phi) = \Psi(0) + \alpha \gamma^{-1} \Delta$ such that $x(t) \leq \Theta$ for all $t \geq 0$.

**Proof** Firstly, $\Theta$ is a functional depending only on $\Phi(t)$, since $\Psi(0) = \int_{-\tau M}^0 \alpha \Phi(s) e^{\gamma s} \, ds$. Then from system (2.2),

$$
\frac{dx(t)}{dt} = \alpha y(t) - y x(t) - \alpha \left[ 1 - \tau' (z(t)) \right] y(t - \tau (z(t))) e^{-\gamma \tau(z(t))}.
$$

Integrating this expression, for $t > 0$, we get

$$
x(t) = e^{-\gamma t} \Psi(0) + \alpha e^{-\gamma t} \int_0^t e^{\gamma s} \left[ y(s) - (1 - \tau'(z(s)) y(s - \tau(z(s)))) \right] \, ds
\leq e^{-\gamma t} \Psi(0) + \alpha e^{-\gamma t} \int_0^t e^{\gamma s} y(s) \, ds
\leq e^{-\gamma t} \Psi(0) + \alpha e^{-\gamma t} \int_0^t e^{\gamma s} \, ds
= e^{-\gamma t} \Psi(0) + \alpha \gamma^{-1} \Delta (1 - e^{-\gamma t}) < \Psi(0) + \alpha \gamma^{-1} \Delta.
$$

Thus, choosing $\Theta(\Phi) = \Psi(0) + \alpha \gamma^{-1} \Delta$, we complete the proof. □

### 3.2 Existence and patterns of equilibria

The purpose of this section is to investigate the existence and patterns of equilibria $(x, y)$ of system (2.2), which satisfy

$$
\begin{align*}
\alpha y - y x - \alpha [1 - \tau'(z) \hat{z}] y e^{-\gamma \tau(z)} &= 0, \\
\alpha [1 - \tau'(z) \hat{z}] y e^{-\gamma \tau(z)} - \beta y^2 &= 0.
\end{align*}
$$

(3.1)

It is clear that system (2.2) has an equilibrium $E_0 = (0, 0)$. 

Theorem 3.5 System (2.2) has exactly one nontrivial equilibrium \( E^\ast = (x^\ast, y^\ast) \), where \( \alpha e^{-\gamma z(x^\ast)} = \alpha e^{-\gamma (x^\ast y^\ast)} = \beta y^\ast \).

Proof Due to \( \dot{z} = \dot{x} + \dot{y} \), if there is a nontrivial equilibrium \( E^\ast = (x^\ast, y^\ast) \), it must satisfy the following equations:

\[
\begin{align*}
\alpha y - \gamma x - \alpha ye^{-\gamma z(x)} &= 0, \\
\alpha ye^{-\gamma z(x)} - \beta y^2 &= 0. \\
\end{align*}
\]  

(3.2)

Adding the two equations of (3.2), we have

\[ x = g(y) = \gamma^{-1}(\alpha y - \beta y^2). \]

It is essential that we need \( 0 < y < \alpha \beta^{-1} \) to make sure that \( x > 0 \). According to Theorem 3.3 and \( \alpha \beta^{-1} e^{-\tau m} < \alpha \beta^{-1} \), we will discuss the existence and uniqueness of nontrivial equilibrium in \( \Lambda = \{ y \in \mathbb{R} | 0 < y < \alpha \beta^{-1} e^{-\tau m} \} \subset \mathbb{R} \). Define \( f : \Lambda \rightarrow \mathbb{R} \) as a continuous mapping by

\[ f(y) = h(y) - \beta y = \alpha e^{-\gamma z(y)} - \beta y, \]

where \( h(y) = \alpha e^{-\gamma z(y)} \) is a bounded function, namely \( \alpha e^{-\gamma z} \leq h(y) \leq \alpha e^{-\gamma z m} \).

Since \( f(0) = \alpha e^{-\gamma z(0)} > 0, f(\alpha \beta^{-1} e^{-\tau m}) = \alpha e^{-\gamma z(\alpha \beta^{-1} e^{-\tau m} + y(\alpha \beta^{-1} e^{-\tau m}))} - \alpha e^{-\gamma z m} < 0 \), this implies that \( f(y) \) has at least one positive zero point \( y^\ast \).

Note that

\[
\begin{align*}
f'(y) &= -\alpha e^{-\gamma z(y)} (y + g(y))(y + \alpha - 2\beta y) - \beta, \\
f''(0) &= -\alpha e^{-\gamma z} \tau'(0)(y + \alpha) - \beta < 0, \\
f''(y) &= \alpha e^{-\gamma z(y)} \left[ \tau'(y + g(y)) \right]^2 (y + \alpha - 2\beta y)^2 + 2\alpha \beta e^{-\gamma z(y)} \tau'(y + g(y)) \\
&\quad - \frac{\alpha}{\gamma} e^{-\gamma z(y)} \tau''(y + g(y))(y + \alpha - 2\beta y)^2 > 0.
\end{align*}
\]

Without loss of generality, assume that \( f(y) \) has two positive roots and the right positive root is denoted as \( y^\ast \). If \( f(y^\ast) \geq 0 \), we have \( f(\alpha \beta^{-1} e^{-\tau m}) > 0 \) due to \( f''(y) > 0 \), giving us a contradiction. Therefore, \( f(y) \) has a unique positive root in the interval \( (0, \alpha \beta^{-1} e^{-\tau m}) \), so system (2.2) has a unique nontrivial equilibrium. The proof is complete. \( \square \)

Remark 3.2 By the proof of Theorem 3.5, we have \( f'(y^\ast) < 0 \), namely

\[ y^\ast \tau'(z^\ast)(2\beta y^\ast - \gamma - \alpha) < 1. \]  

(3.3)

4 Stability of equilibria

In this section, we study the linearized stability of the two equilibria \( E_0 \) and \( E^\ast \) by linearizing system (2.2). Since the delay is a function depending on the state variables \( x \) and \( y \), linearizing an equation with state-dependent delay is not completely straightforward. In 1996, Cooke and Huang [33] solved the linearization problem of state-dependent delay...
differential equations by “freezing the delay” at an equilibrium. We use the same method to linearize system \((2.2)\). To do so, let \(E^*\) be an arbitrary equilibrium and \(x = x^* + x_1\) and \(y = y^* + y_1\). We shall give the details of linearizing the first equation of system \((2.2)\), and linearizing the second could be accomplished similarly. The first equation is linearized as follows:

\[
\dot{x}_1 = \alpha(y^* + y_1 - y(x^* + x_1)) - \alpha \left[ 1 - \lambda'(z^* + x_1 + y_1) \right] y(t - \tau(z^* + x_1 + y_1)) e^{-\gamma \tau(z^* + x_1 + y_1)}
\]

\[
= -\gamma x_1 + \alpha y_1 + \alpha y^* \tau' z^* (\gamma + \alpha - 2 \beta y^*) y_1 e^{-\gamma \tau(z^*)} - \alpha e^{-\gamma \tau(z^*)} y_1(t - \tau(z^*))
\]

\[
= -\gamma x_1 + \left[ \alpha + \alpha y^* \tau' (z^*) (\gamma + \alpha - 2 \beta y^*) \right] y_1 e^{-\gamma \tau(z^*)} y_1(t - \tau(z^*)).
\]

Therefore, the linearized system of \((2.2)\) is

\[
\begin{align*}
\dot{x}_1 &= -\gamma x_1 + \left[ \alpha + \alpha y^* \tau' (z^*) (\gamma + \alpha - 2 \beta y^*) \right] y_1 e^{-\gamma \tau(z^*)} y_1(t - \tau(z^*)), \\
\dot{y}_1 &= -[\alpha y^* \tau' (z^*) (\gamma + \alpha - 2 \beta y^*)] y_1 e^{-\gamma \tau(z^*)} + 2 \beta y^* y_1 e^{-\gamma \tau(z^*)} y_1(t - \tau(z^*)).
\end{align*}
\]

This leads to the following characteristic equation:

\[
(\lambda + \gamma) \left( \lambda + 2 \beta y^* + \theta^* - \alpha e^{-\gamma \tau(z^*)} \right) = 0,
\]

where

\[
\theta^* = \alpha y^* \tau' (z^*) (\gamma + \alpha - 2 \beta y^*) e^{-\gamma \tau(z^*)}.
\]

### 4.1 Linearized stability of the extinction equilibrium \(E_0\)

For the extinction equilibrium \(E_0 = (0,0)\), \((4.1)\) reduces to

\[
(\lambda + \gamma) \left( \lambda - \alpha e^{-\tau(0)(\gamma + \lambda)} \right) = 0.
\]

Clearly, \(\lambda = -\gamma < 0\) is one of these eigenvalues. All the other eigenvalues \(\lambda\) satisfy the equation \(\lambda e^{-\tau(0)(\gamma + \lambda)} = \alpha > 0\), which always has a real, positive solution. Hence \(E_0 = (0,0)\) is a saddle point. We have the following result.

**Theorem 4.1** The extinction equilibrium \(E_0 = (0,0)\) is a saddle point and is unstable.

### 4.2 Linearized stability of the nontrivial equilibrium \(E^*\)

**Theorem 4.2** The nontrivial equilibrium \(E^* = (x^*, y^*)\) is locally asymptotically stable.

**Proof** The characteristic equation

\[
(\lambda + \gamma) \left( \lambda + 2 \beta y^* + \theta^* - \alpha e^{-\gamma \tau(z^*)} \right) = 0,
\]

that is,

\[
(\lambda + \gamma) \left( \lambda - \beta y^* e^{-\gamma \tau(z^*)} + \eta \right) = 0,
\]

where \(\eta = \alpha y^* \tau' (z^*) (\gamma + \alpha - 2 \beta y^*) e^{-\gamma \tau(z^*)}\).
where
\[ \eta = \beta y^* \left[ 2 + y^* \tau'(z^*) (\nu + \alpha - 2\beta y^*) \right]. \]

Obviously, equation (4.2) has a negative real root \( \lambda = -\gamma. \)
Other roots are given by the following equation:
\[ G(\lambda) = \lambda - \beta y^* e^{-\tau(z^*)} + \eta = 0. \]  (4.3)

By inequality (3.3) of Remark 3.2, we have \( G(0) = \eta - \beta y^* > 0. \) Hence, zero is not the root of equation (4.3).
Now, let us prove that equation (4.3) has no purely imaginary roots.
Assume that equation (4.3) has a purely imaginary root \( \lambda = iv, \) where \( v > 0. \) Substituting it into equation (4.3) and separating the real and the imaginary parts, we obtain
\[
\begin{align*}
v &= -\beta y^* \sin(\tau(z^*)v), \\
\eta &= \beta y^* \cos(\tau(z^*)v).
\end{align*}
\]

Given \((\sin(\tau(z^*)v))^2 + (\cos(\tau(z^*)v))^2 = 1\) and inequality (3.3), we have
\[
v^2 = (\beta y^*)^2 - \eta^2
= (\beta y^*)^2 - \left( \beta y^* \right)^2 \left[ 2 + y^* \tau'(z^*) (\nu + \alpha - 2\beta y^*) \right]^2
= (\beta y^*)^2 \left[ 1 - (2 + y^* \tau'(z^*) (\nu + \alpha - 2\beta y^*))^2 \right]
< 0,
\]
which is a contradiction. Therefore, equation (4.3) has no purely imaginary roots, and each root of the characteristic equation has a negative real part. The proof is complete. \( \Box \)

5 Permanence
In this section, we will discuss the permanence of system (2.2) by using the method of [5, 28], and obtain explicit bounds for the eventual behaviors of \( x(t) \) and \( y(t) \), which are independent of the initial conditions.

Definition 5.1 System (2.2) is said to be permanent if there exist positive constants \( m, n, M, \) and \( N \) such that every positive solution \((x(t), y(t))\) of system (2.2) satisfies
\[
m \leq \liminf_{t \to \infty} x(t) \leq \limsup_{t \to \infty} x(t) \leq M,
\]
\[
n \leq \liminf_{t \to \infty} y(t) \leq \limsup_{t \to \infty} y(t) \leq N.
\]

In order to obtain explicit bounds for the eventual behavior of \( y(t) \), we first investigate the upper and lower bounds of \( \lim_{t \to \infty} y(t) \) in the case where \( y(t) \) is eventually monotone.

Theorem 5.1 Suppose that \( y(t) \) is eventually monotone. Then
\[
\alpha \beta^{-1} e^{-\gamma M} \leq \lim_{t \to \infty} y(t) \leq \alpha \beta^{-1} e^{-\nu M}.
\]
Proof Since \( y(t) \) is eventually monotone and bounded, there exists a \( 0 \leq \bar{y} < \infty \) such that
\[
\lim_{t \to \infty} y(t) = \bar{y} \quad \text{and} \quad \lim_{t \to \infty} \dot{y}(t) = 0.
\]
Because the zero solution is unstable by Theorem 4.1, we obtain \( \bar{y} \neq 0 \), namely \( 0 < \bar{y} < \infty \). For the boundedness of \( z(t) \) and \( \dot{z}(t) = -\gamma z(t) + (\alpha + \gamma)y(t) - \beta z^2(t) \), according to the Barbálat lemma [34], we have \( \lim_{t \to \infty} \dot{z}(t) = 0 \). Thus, from system (2.2), taking the limit superior as \( t \to \infty \), we get that
\[
0 \leq \bar{y} \left[ e^{-\gamma \lim_{t \to \infty} \tau(t)} - \beta \bar{y} \right].
\]
Hence \( \bar{y} \leq \alpha \beta^{-1} e^{-\gamma \lim_{t \to \infty} \tau(t)} \leq \alpha \beta^{-1} e^{-\gamma \tau_m} \) by combining with the hypotheses (A2).

Similarly, from system (2.2), taking the limit infimum as \( t \to \infty \), we obtain that
\[
0 \geq \bar{y} \left[ e^{-\gamma \lim_{t \to \infty} \tau(t)} - \beta \bar{y} \right].
\]
Thus \( \bar{y} \geq \alpha \beta^{-1} e^{-\gamma \lim_{t \to \infty} \tau(t)} \geq \alpha \beta^{-1} e^{-\gamma \tau_m} \) by the hypotheses (A2). The proof is complete.

Now let us show explicit bounds, independent of initial conditions, for \( y(t) \).

**Theorem 5.2** There exist positive constants \( n \) and \( N \) such that, for any positive solution \((x(t), y(t))\) of model (2.2),
\[
n \leq \liminf_{t \to \infty} y(t) \leq \limsup_{t \to \infty} y(t) \leq N,
\]
where \( n = \alpha \beta^{-1} e^{-\gamma \tau_m} \), \( N = \alpha \beta^{-1} e^{-\gamma \tau_m} \).

**Proof** Our proof is split into two cases.

In the first case, \( y(t) \) is eventually monotonic. According to Theorem 5.1 and the definitions of superior and inferior limits, the claim is true.

Secondly, we consider an oscillatory \( y(t) \). We only show that \( \limsup_{t \to \infty} y(t) \leq \alpha \beta^{-1} e^{-\gamma \tau_m} \), because the other inequality can be obtained analogously. Define the sequence \( \{t_m\} \) as those times for which \( y(t) \) obtains its local maximum, i.e., \( \dot{y}(t_m) = 0 \), \( y(t_m) < 0 \).

Let \( \bar{y} = \limsup_{t \to \infty} y(t) \). Then \( \bar{y} = \limsup_{t \to \infty} y(t) \). If \( \bar{y} \leq \alpha \beta^{-1} e^{-\gamma \tau_m} \), we are done. Therefore, we assume that
\[
\bar{y} > \alpha \beta^{-1} e^{-\gamma \tau_m}. \tag{5.1}
\]
Then \( 0 = \dot{y}(t_m) = \alpha \left[ 1 - \tau(z_m) \right] y(t - \tau(z_m)) e^{-\gamma \tau(t_m)} - \beta y^2(t_m), \) where \( z_m = z(t_m) = x(t_m) + y(t_m) \).

Now let us choose a subsequence of \( \{t_m\} \), labeled as \( \{t_{m_1}\} \), so that \( \lim_{m_1 \to \infty} y(t_{m_1}) = \bar{y} \) and \( \limsup_{m_1 \to \infty} z(t_{m_1}) = \bar{z} \).

We then choose a subsequence of \( \{t_{m_1}\} \), labeled as \( \{t_{m_2}\} \), so that \( \lim_{m_2 \to \infty} z(t_{m_2}) = \bar{z} \) and \( \limsup_{m_2 \to \infty} y(t_{m_2} - \tau(z_{m_2})) = \bar{y} \).

Finally, we choose a subsequence of \( \{t_{m_2}\} \), labeled as \( \{t_{m_3}\} \), so that \( \lim_{m_3 \to \infty} y(t_{m_3} - \tau(z_{m_3})) = \bar{y} \).
Then from (2.2) and (5.1), taking the limit as \( m_1 \to \infty, m_2 \to \infty, m_3 \to \infty, \)

\[
0 = \alpha [1 - \tau'(\hat{z})\hat{y} e^{-\gamma \tau(\hat{z})} - \beta \hat{y}^2] \\
< \alpha [1 - \tau'(\hat{z})\hat{y} e^{-\gamma \tau(\hat{z})} - \beta \hat{y}] .
\]

If \( \tilde{y} \leq \hat{y} \), we have a contradiction. Therefore, assuming \( \tilde{y} > \hat{y} \), we can choose a value \( t_n \) for \( m, m_1, m_2, m_3 \) such that \( \dot{y}(t_n) = 0, \hat{y}(t_n) < 0, \) and \( \lim sup_{n \to \infty} y(t_n) \geq \bar{y} > \hat{y} \). However, this contradicts the definition of \( \tilde{y} \), therefore \( \tilde{y} > \hat{y} \) cannot be true, which eliminates the last possibility. The proof is complete. 

Now let us give estimates for \( x \) by using the estimates obtained in Theorem 5.1. It should be noted that we can find a \( T(\epsilon) > 0 \) so large that

\[
\alpha \beta^{-1} e^{-\gamma T} - \epsilon \leq y(t) \leq \alpha \beta^{-1} e^{-\gamma T} + \epsilon \quad (5.2)
\]

for any given \( \epsilon > 0 \) whenever \( t \geq T \). Clearly, inequality (5.2) is also valid for those \( t \geq T + \tau_M \).

**Theorem 5.3** There exist positive constants \( m \) and \( M \) such that, for any positive solution \((x(t), y(t))\) of model (2.2),

\[
m \leq \lim inf_{t \to \infty} x(t) \leq \lim sup_{t \to \infty} x(t) \leq M,
\]

where \( m = \alpha^2 \gamma^{-1} \beta^{-1} e^{-\gamma T} (1 - e^{-\gamma T}) , M = \alpha^2 \gamma^{-1} \beta^{-1} e^{-\gamma T} (1 - e^{-\gamma T}) \).

**Proof** Since the first equation of system (2.2) can be written in the integral equation form and utilizing inequality (5.2), we have for \( t \geq T + \tau_M \),

\[
x(t) = \int_{t-\tau(z(t))}^{t} \alpha y(\xi) e^{-\gamma (t-\xi)} d\xi \\
\leq \int_{t-\tau(z(t))}^{t} \alpha (\alpha \beta^{-1} e^{-\gamma T} + \epsilon) e^{-\gamma (t-\xi)} d\xi \\
= \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma T} + \epsilon) \left[ 1 - e^{-\gamma T} \right],
\]

where \( \epsilon > 0 \) is arbitrary. Thus we have

\[
\lim sup_{t \to \infty} x(t) \leq \lim sup_{t \to \infty} \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma T} + \epsilon) \left[ 1 - e^{-\gamma T} \right] \\
\leq \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma T} + \epsilon) \left( 1 - e^{-\gamma T} \right) \\
= \alpha^2 \gamma^{-1} \beta^{-1} e^{-\gamma T} (1 - e^{-\gamma T}) + \epsilon \alpha \gamma^{-1} (1 - e^{-\gamma T}).
\]

Since \( \epsilon \) is arbitrary, we get an upper bound on \( x(t) \).
Similarly, we have for \( t \geq T + \tau_M \) that

\[
x(t) \geq \int_{t-\tau(z(t))}^{t} \alpha (\alpha \beta^{-1} e^{-\gamma r M} - \epsilon) e^{-\gamma \xi} d\xi
\]

\[
= \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma r M} - \epsilon) \left[ 1 - e^{-\gamma \tau(z(t))} \right],
\]

where \( \epsilon > 0 \) is arbitrary. Then we obtain

\[
\lim \inf_{t \to \infty} x(t) \geq \lim \inf_{t \to \infty} \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma r M} - \epsilon) \left[ 1 - e^{-\gamma \tau(z(t))} \right]
\]

\[
\geq \alpha \gamma^{-1} (\alpha \beta^{-1} e^{-\gamma r M} - \epsilon) (1 - e^{-\gamma r m})
\]

\[
= \alpha^2 \gamma^{-1} \beta^{-1} e^{-\gamma r M} (1 - e^{-\gamma r m}) - \epsilon \alpha \gamma^{-1} (1 - e^{-\gamma r m}).
\]

Therefore, we get a lower bound on \( x(t) \) by the arbitrariness of \( \epsilon \). The proof is complete. \( \square \)

**Remark 5.1** In Theorem 5.6 of [14] and Theorem 5.4 of [28], one requires that \( \tau_M < 2 \tau_m \) for the lower bound to be positive. In fact, according to Theorem 5.3, we see that this condition is not necessary.

### 6 Conclusions and discussions

In this paper, based on the biological observations that during World War II the matura-
tion time of seals and whales was not a fixed value, but depended on the total population
(immature and mature), starting with an age-structured model (2.1), we formulated and
analyzed a novel stage-structured model with state-dependent maturity delay.

Compared with the previous state-dependent delay equations (e.g., [5, 26, 28, 30]),
model (2.2) is not directly changing the constant delay \( \tau \) into a state-dependent delay \( \tau(z(t)) \) but was obtained by reducing the age-structured population model, which has the
correction term \( 1 - \tau'(z(t)) \check{z}(t) \). Biologically speaking, model (2.2) is appropriate in terms of population modeling. On the one hand, with the state-dependent maturity delay, the
changes in the number of mature individuals depend on reproduction and death and the
changing definition of maturity, which is in line with the correction term \( 1 - \tau'(z(t)) \check{z}(t) \).

On the other hand, we can represent \( x(t) \) in an integral form by some biological inductions, namely

\[
x(t) = \int_{t-\tau(z(t))}^{t} \alpha y(\xi) e^{-\gamma (t-\xi)} d\xi.
\]

Taking the derivatives of \( x(t) \), we obtain the first equation of model (2.2).

From a biological point of view, we show that \( t - \tau(z(t)) \) should be a strictly increasing
function of \( t \) without any conditions and the derivative with respect to time of the state-dependent maturity delay \( \tau(z(t)) \) is strictly less than one. In addition, it is biologically rea-
sonable for the assumption of the delay \( \tau(z(t)) \). The biological phenomenon mentioned
above, a non-decreasing delay, implies that a larger population leads to a longer develop-
mental duration, and it makes clear the stabilizing effect [15, 35].

Mathematically compared with [5, 28], first of all, the positivity and boundedness of
solutions are discussed, which do not need the stringent condition on \( \tau'(z(t)) \) to ensure
the positivity of $x$. Then the existence and uniqueness of all equilibria are formulated, which are different from the result of [5]. Besides, the nontrivial equilibrium $E^*$ is always linearly stable without auxiliary conditions. Finally, we discuss some global properties of the solutions of our model (2.2) and obtain explicit bounds for the eventual behaviors of $x(t)$ and $y(t)$, which do not need the condition $\tau_M < 2\tau_m$ for the lower bound of $x(t)$.

Further research in this direction may consider more realistic complex models, for example, a multi-stage structure population model with state-dependent delays and a multi-population model with state-dependent delays.

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Authors’ contributions
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Author details
1 Key Laboratory of Eco-environments in Three Gorges Reservoir Region (Ministry of Education), School of Mathematics and Statistics, Southwest University, Chongqing, China. 2 School of Mathematics and Statistics, Guangxi Teachers Education University, Nanning, China.

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