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Analysis of a predator-prey model with impulsive diffusion and releasing on predator population

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Abstract

In this paper, we establish a predator-prey model with impulsive diffusion and releasing on predator population. This predator-prey model for two regions, which are connected by diffusion of predator population, portrays the evolvement of population. We prove that all solutions of the investigated system are uniformly ultimately bounded. We also prove that there exists globally asymptotically stable prey-extinction boundary periodic solution. The condition for permanence is obtained. Simulations are also employed to verify our results. It is discovered that the increasing diffusive rate of predator population will count against the pest management. We conclude that the impulsive diffusion and releasing predator provide reliable tactic basis for pest management.

Keywords: predator-prey model; impulsive diffusion; impulsive releasing; extinction; permanence

1 Introduction

The warfare between human and pests has sustained for thousands of years. In the past few decades, man have adopted some advanced and modern weapons for instance chemical pesticides, biological pesticides, remote sensing and measure, computers, atomic energy, *et cetera.* Some brilliant achievements have been obtained. However, the warfare will never be over. Although a great number of pesticides were used to control pests, the insect pests impairing crops are increasing for the resistance to pesticides. With pesticides employed, the residual pests breed a large number of pests with resistance to pesticides. So the pesticide is invalid in some sense. Moreover, insect pests will continue. On the other hand, the chemical pesticides kill not only pests but also their natural enemies. Therefore, insect pests are rampant again. Then the effect of chemical control may give rise to disastrous results, for example, environmental contamination and toxicosis of the man and animals and so on.

The use of natural enemy to suppress pests is one of the most important approaches in pest control. Biological control [1-8] is one of the reduction in pest populations from the actions of other living organisms, often called natural enemies or beneficial species. It is the purposeful introduction and establishment of one or more natural enemies from



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region of origin of an exotic pest, specifically for the purpose of suppressing the abundance of the pest in a new target region to a level at which it no longer causes economic damage. Jiao *et al.* [9] analyzed the dynamics of a stage-structured Holling mass defence predator-prey model with impulsive perturbations on predators

$$\begin{cases} x_1'(t) = rx_2(t) - re^{-w\tau_1}x_2(t - \tau_1) - wx_1(t), \\ x_2'(t) = re^{-w\tau_1}x_2(t - \tau_1) - \frac{\beta x_2(t)}{1 + ax_2 + bx_2^2}x_3(t) - d_3x_2(t) - d_4x_2^2(t), \\ x_3'(t) = \frac{k\beta x_2(t)}{1 + ax_2 + bx_2^2}x_3(t) - dx_3(t), \\ \Delta x_1(t) = 0, \\ \Delta x_2(t) = 0, \\ \Delta x_3(t) = \mu, \\ (\varphi_1(\zeta), \varphi_2(\zeta), \varphi_3(\zeta)) \in C_+ = C([-\tau_1, 0], R_+^3), \quad \varphi_i(0) > 0, i = 1, 2, 3, \end{cases}$$
(1)

where $x_1(t)$ and $x_2(t)$ represent the immature and mature pest densities, respectively, and $x_3(t)$ denotes the density of nature enemy. The biological meanings of parameters can be seen in reference [9].

The dispersal is a ubiquitous phenomenon in the natural world. It is important for us to understand the ecological and evolutionary dynamics of populations mirrored by the large number of mathematical models devoted to it in the scientific literature [10-13]. In recent years, the analysis of these models focus on the coexistence of population and local (or global) stability of equilibria [14-20]. Spatial factors play a fundamental role in the persistence and stability of the population, although the complete results have not yet been obtained even in the simplest one-species case. Whereas the population dynamics with the effects of spatial heterogeneity is modeled by a diffusion process, most previous papers focused on the population dynamical system modeled by the ordinary differential equations. But in practice, it is often the case that diffusion occurs in regular pulse. For example, when winter comes, birds migrate between patches in search for a better environment, whereas they do not diffuse in other seasons, and the excursion of foliage seeds occurs at fixed period of time every year. Thus, impulsive diffusion provides a more natural description. Lately theories of impulsive differential equations [21] have been introduced into population dynamics. Jiao et al. [22] propose to investigate the dynamical behaviors of a stage-structured predator-prey model with prey impulsively diffusing between two patches

$$\begin{cases} \frac{dx_{1}(t)}{dt} = x_{1}(t)(a_{1} - b_{1}x_{1}(t)), \\ \frac{dx_{2}(t)}{dt} = x_{2}(t)(a_{2} - b_{2}x_{2}(t)) - \alpha x_{2}(t)y_{2}(t), \\ \frac{dy_{1}(t)}{dt} = k\alpha x_{2}(t)y_{2}(t) - k\alpha e^{-w\tau_{1}}x_{2}(t - \tau_{1})y_{2}(t - \tau_{1}) - wy_{1}(t), \\ \frac{dy_{2}(t)}{dt} = k\alpha e^{-w\tau_{1}}x_{2}(t - \tau_{1})y_{2}(t - \tau_{1}) - dy_{2}(t), \\ \Delta x_{1}(t) = d_{1}(x_{2}(t) - x_{1}(t)), \\ \Delta x_{2}(t) = d_{2}(x_{1}(t) - x_{2}(t)), \\ \Delta y_{1}(t) = 0, \\ \Delta y_{2}(t) = 0, \end{cases}$$

$$t = n\tau, n = 1, 2, \dots,$$

$$(2)$$

where we suppose that the system is composed of two patches connected by diffusion and occupied by a single species x_i (i = 1, 2) is the density of species in the *i*th patch, and $y_1(t)$ and $y_2(t)$ represent the densities of the immature individual predator and mature individual predator at time t in the second patch. The biological meanings of parameters can be seen in reference [22].

Theories of impulsive differential equations have been introduced into population dynamics lately [23–28]. Impulsive equations are found in almost every domain of applied science and have been studied in many investigations [28–33]; they generally describe phenomena that are subject to steep or instantaneous changes. The theories of population dynamical systems and their applications have achieved many good results. In this paper, we investigate a predator-prey model with impulsive diffusion and releasing on predator population. We expect to obtain some dynamical properties of the investigated system. We also expect that the impulsive diffusion and releasing predator will provide reliable tactic basis for pest management.

The organization of this paper is as follows. In the next section, we introduce the model and background concepts. In Section 3, some important lemmas are presented. We give the globally asymptotically stable conditions of the prey-extinction boundary periodic solution of system (3) and the permanent condition of system (3) in Section 4. Simulation analysis and brief discussion are given in the last section to conclude this work.

2 The model

In this paper, we establish a predator-prey model with impulsive diffusion and releasing on predator population:

$$\begin{cases} \frac{dx_{1}(t)}{dt} = x_{1}(t)(a_{1} - b_{1}x_{1}(t)) - \frac{\beta_{1}x_{1}(t)y_{1}(t)}{\sigma_{1} + x_{1}(t)},\\ \frac{dy_{1}(t)}{dt} = \frac{k_{1}\beta_{1}x_{1}(t)y_{1}(t)}{\sigma_{1} + x_{1}(t)} - d_{1}y_{1}(t),\\ \frac{dx_{2}(t)}{dt} = x_{2}(t)(a_{2} - b_{2}x_{2}(t)) - \frac{\beta_{2}x_{2}(t)y_{2}(t)}{\sigma_{2} + x_{2}(t)},\\ \frac{dy_{2}(t)}{dt} = \frac{k_{2}\beta_{2}x_{2}(t)y_{2}(t)}{\sigma_{2} + x_{2}(t)} - d_{2}y_{2}(t),\\ \Delta x_{1}(t) = 0,\\ \Delta y_{1}(t) = D(y_{2}(t) - y_{1}(t)),\\ \Delta x_{2}(t) = 0,\\ \Delta y_{2}(t) = D(y_{1}(t) - y_{2}(t)), \end{cases} t = (n + l)\tau, n \in Z_{+},$$

$$(3)$$

$$\begin{cases} \Delta y_{1}(t) = 0,\\ \Delta y_{1}(t) = \mu_{1},\\ \Delta x_{2}(t) = 0,\\ \Delta y_{2}(t) = \mu_{2}, \end{cases} t = (n + 1)\tau, n \in Z_{+},$$

where we suppose that the system is composed of two patches connected by diffusion. These two patches are separated by rivers or highways or railways. The predator population can traverse the rivers or highways or railways, whereas the prey population not. In this system, $x_i(t)$ and $y_i(t)$ represent the numbers of prey and predator populations in patch i (i = 1, 2) at time $t, a_i > 0$ represents the intrinsic growth rate of the prey population in patch i (i = 1, 2), and $b_i > 0$ represents the coefficient of the intraspecific competition of the prey population in patch i (i = 1, 2). The predator consumes the prey according to Holling type-II functional response

$$\frac{\beta_i x_i(t)}{\sigma_i + x_i(t)} \quad (i = 1, 2)$$

with the half-saturation constant σ_i in patch i (i = 1, 2) at time t. k_i (i = 1, 2) is the rate of conversion of nutrients into the reproduction of the predator in patch i (i = 1, 2), d_i (i = 1, 2) represents the death in patch i (i = 1, 2). The pulse diffusion occurs every τ period (τ is a positive constant), the system evolves from its initial state without being further affected by diffusion until the next pulse appears; $\Delta y_i((n + l)\tau) = y_i((n + l)\tau^+) - y_i((n + l)\tau)$, where $y_i((n + l)\tau^+)$ represents the density of population in the *i*th patch immediately after the *n*th diffusion pulse at time $t = (n + l)\tau$, whereas $y_i((n + l)\tau)$ represents the density of population in the *i*th patch before the *n*th diffusion pulse at time $t = (n + l)\tau$, $0 < l < 1, n \in Z_+, 0 < D < 1$ represents the diffusive rate between the patches, $\Delta y_i((n + 1)\tau) = y_i((n + 1)\tau^+) - y_i((n + 1)\tau)$, and μ_i (i = 1, 2) represents the releasing amount of predator population at $t = (n + 1)\tau$, $n \in Z_+$ in patch i (i = 1, 2).

3 The lemmas

The solution of (3), denoted by $X(t) = (x_1(t), y_1(t), x_2(t), y_2(t))^T$, is a piecewise continuous function $X : R_+ \to R_+^4$, X(t) is continuous on $(n\tau, (n+l)\tau]$ and $((n+l)\tau, (n+1)\tau]$, $n \in Z_+$, and $X(n\tau^+) = \lim_{t\to n\tau^+} X(t)$, $X((n+l)\tau^+) = \lim_{t\to (n+l)\tau^+} X(t)$ exist. Obviously, the global existence and uniqueness of solutions of (3) is guaranteed by the smoothness properties of f, the mapping defined by the right side of system (3) [21].

- Let $V: \mathbb{R}_+ \times \mathbb{R}_+^4 \to \mathbb{R}_+$. Then V is said to belong to class V_0 if
 - (i) V is continuous in (nτ, (n + l)τ] × R⁴₊ and ((n + l)τ, (n + 1)τ] × R⁴₊ for all z ∈ R⁴₊, n ∈ Z₊, and V(nτ⁺, z) = lim_{(t,y)→(nτ⁺,z)} V(t, y) and V((n + l)τ⁺, z) = lim_{(t,y)→((n+l)τ⁺,y)} V(t, y) exist;
- (ii) V is locally Lipschitzian in z.

Definition 3.1 If $V \in V_0$, then, for $(t, z) \in (n\tau, (n + l)\tau] \times R^4_+$ and $((n + l)\tau, (n + 1)\tau] \times R^4_+$, the upper right derivative of V(t, z) with respect to the impulsive differential system (3) is defined as

$$D^{+}V(t,z) = \limsup_{h \to 0} \frac{1}{h} \Big[V \big(t + h, z + hf(t,z) \big) - V(t,z) \Big].$$

Since $\frac{dx_i(t)}{dt} = 0$ when $x_i(t) = 0$, $\frac{dy_i(t)}{dt} = 0$ when $y_i(t) = 0$, and $\Delta y_i(t) = \mu_i > 0$ when $t = (n+1)\tau$, we easily obtain the following lemma.

Lemma 3.2 Suppose that X(t) is a solution of (3) with $X(0^+) \ge 0$. Then $X(t) \ge 0$ for $t \ge 0$, and further X(t) > 0 ($t \ge 0$) for $X(0^+) > 0$.

Lemma 3.3 [21] Let the function $m \in PC'[R^+, R]$ satisfy the inequalities

$$\begin{cases} m'(t) \le p(t)m(t) + q(t), & t \ge t_0, t \ne t_k, k = 1, 2, \dots, \\ m(t_k^+) \le d_k m(t_k) + b_k, & t = t_k, \end{cases}$$
(4)

where $p, q \in C[R^+, R]$, and $d_k \ge 0$ and b_k are constants. Then

$$m(t) \le m(t_0) \prod_{t_0 < t_k < t} d_k \exp\left(\int_{t_0}^t p(s) \, ds\right) + \sum_{t_0 < t_k < t} \left(\prod_{t_k < t_j < t} d_j \exp\left(\int_{t_k}^t p(s) \, ds\right)\right) b_k$$
$$+ \int_{t_0}^t \prod_{s < t_k < t} d_k \exp\left(\int_s^t p(\sigma) \, d\sigma\right) q(s) \, ds, \quad t \ge t_0.$$

Now, we show that all solutions of (3) are uniformly ultimately bounded.

Lemma 3.4 There exists a constant M > 0 such that $x_i(t) \le M$, $y_i(t) \le M$ (i = 1, 2) for each solution $(x_1(t), y_1(t), x_2(t), y_2(t))$ of (3) with all t large enough.

Proof Define

$$V(t) = k_1 x_1(t) + y_1(t) + k_2 x_2(t) + y_2(t),$$

and $\lambda = \min_{i=1,2} \{d_i\}$. When $t \neq n\tau$, $t \neq (n + l)\tau$, we have

$$D^{+}V(t) + \lambda V(t)$$

$$= k_{1}x_{1}(t)[(a_{1} + \lambda) - b_{1}x_{1}(t)] - (d_{1} - \lambda)y_{1}(t)$$

$$+ k_{2}x_{2}(t)[(a_{2} + \lambda) - b_{2}x_{2}(t)] - (d_{2} - \lambda)y_{2}(t)$$

$$\leq k_{1}x_{1}(t)[(a_{1} + \lambda) - b_{1}x_{1}(t)] + k_{2}x_{2}(t)[(a_{2} + \lambda) - b_{2}x_{2}(t)]$$

$$= -k_{1}b_{1}\left(x_{1}(t) - \frac{a_{1} + \lambda}{2b_{1}}\right)^{2} + \frac{k_{1}(a_{1} + \lambda)^{2}}{4b_{1}}$$

$$- k_{2}b_{2}\left(x_{2}(t) - \frac{a_{2} + \lambda}{2b_{2}}\right)^{2} + \frac{k_{2}(a_{2} + \lambda)^{2}}{4b_{2}}$$

$$\leq \frac{k_{1}(a_{1} + \lambda)^{2}}{4b_{1}} + \frac{k_{2}(a_{2} + \lambda)^{2}}{4b_{2}} \triangleq \zeta.$$

When $t = n\tau$, we have

$$V(n\tau^{+}) = k_1 x_1(n\tau^{+}) + y_1(n\tau^{+}) + k_2 x_2(n\tau^{+}) + y_2(n\tau^{+})$$

= $k_1 x_1(n\tau) + y_1(n\tau) + \mu_1 + k_2 x_2(n\tau) + y_2(n\tau) + \mu_2$
= $k_1 x_1(n\tau) + y_1(n\tau) + k_2 x_2(n\tau) + y_2(n\tau) + \mu_1 + \mu_2$
= $V(n\tau) + (\mu_1 + \mu_2).$

When $t = (n + l)\tau$, we have

$$V((n+l)\tau^{+}) = k_{1}x_{1}((n+l)\tau^{+}) + y_{1}((n+l)\tau^{+}) + k_{2}x_{2}((n+l)\tau^{+}) + y_{2}((n+l)\tau^{+})$$

$$= k_{1}x_{1}((n+l)\tau) + (1-D)y_{1}((n+l)\tau) + Dy_{2}((n+l)\tau) + k_{2}x_{2}((n+l)\tau)$$

$$+ Dy_{1}((n+l)\tau) + (1-D)y_{2}((n+l)\tau)$$

$$= k_{1}x_{1}((n+l)\tau) + y_{1}((n+l)\tau) + k_{2}x_{2}((n+l)\tau) + y_{2}((n+l)\tau)$$

$$= V((n+l)\tau).$$

By Lemma 3.3, for $t \in (n\tau, (n+1)\tau]$, we have

$$V(t) \leq V(0^{+})e^{-\lambda t} + \frac{\zeta}{\lambda}(1 - e^{-\lambda t}) + (\mu_{1} + \mu_{2})\frac{e^{-\lambda(t-\tau)}}{1 - e^{\lambda\tau}} + (\mu_{1} + \mu_{2})\frac{e^{\lambda\tau}}{e^{\lambda\tau} - 1}$$

$$\rightarrow \frac{\zeta}{\lambda} + (\mu_{1} + \mu_{2})\frac{e^{\lambda\tau}}{e^{\lambda\tau} - 1} \quad \text{as } t \rightarrow \infty.$$

So V(t) is uniformly ultimately bounded. Hence, by the definition of V(t) we have that there exists a constant M > 0 such that $x_i(t) \le M$, $y_i(t) \le M$ (i = 1, 2) for t large enough. The proof is complete.

If $x_i(t) = 0$ (*i* = 1, 2), then we have the subsystem of (3)

$$\begin{cases} \frac{dy_{1}(t)}{dt} = -d_{1}y_{1}(t), \\ \frac{dy_{2}(t)}{dt} = -d_{2}y_{2}(t), \end{cases} t \neq (n+l)\tau, t \neq (n+1)\tau, \\ \Delta y_{1}(t) = D(y_{2}(t) - y_{1}(t)), \\ \Delta y_{2}(t) = D(y_{1}(t) - y_{2}(t)), \end{cases} t = (n+l)\tau,$$

$$\begin{cases} \Delta y_{1}(t) = -\mu_{1}, \\ \Delta y_{2}(t) = -\mu_{2}, \end{cases} t = (n+1)\tau, n = 1, 2, \dots. \end{cases}$$
(5)

We easily obtain the analytic solution of (5) between pulses as follows:

$$\begin{cases} y_{1}(t) = \begin{cases} y_{1}(n\tau^{+})e^{-d_{1}(t-n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{1}((n+l)\tau^{+})e^{-d_{1}(t-(n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \end{cases} \\ y_{2}(t) = \begin{cases} y_{2}(n\tau^{+})e^{-d_{2}(t-n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{2}((n+l)\tau^{+})e^{-d_{2}(t-(n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau]. \end{cases} \end{cases}$$
(6)

Considering the third and fourth equations of (5), we have

$$\begin{cases} y_1((n+l)\tau^+) = (1-D)e^{-d_1l\tau}y_1(n\tau^+) + De^{-d_2l\tau}y_2(n\tau^+), \\ y_2((n+l)\tau^+) = De^{-d_1l\tau}y_1(n\tau^+) + (1-D)e^{-d_2l\tau}y_2(n\tau^+). \end{cases}$$
(7)

Considering the fifth and sixth equations of (5), we also have

$$\begin{cases} y_1((n+1)\tau^+) = y_1((n+l)\tau^+)e^{-d_1(1-l)\tau} + \mu_1, \\ y_2((n+1)\tau^+) = y_2((n+l)\tau^+)e^{-d_2(1-l)\tau} + \mu_2. \end{cases}$$
(8)

Substituting (7) into (8), we have the stroboscopic map of (5)

$$\begin{cases} y_1((n+1)\tau^+) = (1-D)e^{-d_1\tau}y_1(n\tau^+) + De^{-[d_1(1-l)+d_2l]\tau}y_2(n\tau^+) + \mu_1, \\ y_2((n+1)\tau^+) = De^{-[d_1l+d_2(1-l)]\tau}y_1(n\tau^+) + (1-D)e^{-d_2\tau}y_2(n\tau^+) + \mu_2. \end{cases}$$
(9)

System (9) has one fixed point

$$\begin{cases} y_1^* = \frac{\mu_2 B_1 + \mu_1 (1 - B_2)}{(1 - A_1) (1 - B_2) - A_2 B_1} > 0, \\ y_2^* = \frac{\mu_1 A_2 + \mu_2 (1 - A_1)}{(1 - A_1) (1 - B_2) - A_2 B_1} > 0, \end{cases}$$
(10)

where

$$\begin{split} A_1 &= (1-D)e^{-d_1\tau} < 1, \\ B_1 &= De^{-[d_1(1-l)+d_2l]\tau} < 1, \\ A_2 &= De^{-[d_1l+d_2(1-l)]\tau} < 1, \\ B_2 &= (1-D)e^{-d_2\tau} < 1. \end{split}$$

Lemma 3.5 The fixed point (y_1^*, y_2^*) of (9) is globally asymptotically stable.

Proof For convenience, we denote $(y_1^n, y_2^n) = (y_1(n\tau^+), y_2(n\tau^+))$. The linear form of (9) can be written as

$$\begin{pmatrix} y_1^{n+1} \\ y_2^{n+1} \end{pmatrix} = M \begin{pmatrix} y_1^n \\ y_2^n \end{pmatrix}.$$
(11)

Obviously, the near dynamics of (y_1^*, y_2^*) is determined by linear system (11). The stability of (y_1^*, y_2^*) is determined by the eigenvalue of *M* less than 1. If *M* satisfies the *Jury* criteria [34], then we know that the eigenvalue of *M* is less than 1,

 $1 - \operatorname{tr} M + \det M > 0.$ (12)

We easily see that (y_1^*, y_2^*) is a unique fixed point of (9) and

$$M = \begin{pmatrix} A_1 & B_1 \\ A_2 & B_2 \end{pmatrix}.$$
 (13)

Since

$$\begin{split} 1 &-\operatorname{tr} M + \det M \\ &= 1 - (A_1 + B_2) + (A_1 B_2 - A_2 B_1) \\ &= (1 - A_1)(1 - B_2) - A_2 B_1 \\ &= \left[1 - (1 - D)e^{-d_1 \tau} \right] \times \left[1 - (1 - D)e^{-d_2 \tau} \right] - D^2 e^{-(d_1 + d_2) \tau} \\ &= \left[\left(1 - e^{-d_1 \tau} \right) + D e^{-d_1 \tau} \right] \left[\left(1 - e^{-d_2 \tau} \right) + D e^{-d_2 \tau} \right] - D^2 e^{-(d_1 + d_2) \tau} \\ &= (1 - e^{-d_1 \tau}) \times \left(1 - e^{-d_2 \tau} \right) + D e^{-d_2 \tau} \left(1 - e^{-d_1 \tau} \right) + D e^{-d_1 \tau} \left(1 - e^{-d_2 \tau} \right) \\ &> 0, \end{split}$$

by the *Jury* criteria, (y_1^*, y_2^*) is locally stable, and then, it is globally asymptotically stable. This completes the proof.

Lemma 3.6 The periodic solution $(\widetilde{y_1(t)}, \widetilde{y_2(t)})$ of system (5) is globally asymptotically stable, where

$$\begin{cases} \widetilde{y_1(t)} = \begin{cases} y_1^* e^{-d_1(t-n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_1^{**} e^{-d_1(t-(n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \\ \widetilde{y_2(t)} = \begin{cases} y_2^* e^{-d_2(t-n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_2^{**} e^{-d_2(t-(n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \end{cases} \end{cases}$$
(14)

where y_1^* and y_2^* are determined as in (10), and y_1^{**} and y_2^{**} are defined as

$$\begin{cases} y_1^{**} = (1-D)e^{-d_1 l\tau} y_1^* + De^{-d_2 l\tau} y_2^*, \\ y_2^{**} = De^{-d_1 l\tau} y_1^* + (1-D)e^{-d_2 l\tau} y_2^*. \end{cases}$$
(15)

4 The dynamics

Theorem 4.1 If

$$D < \frac{1}{2} \tag{16}$$

and

$$\max_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [\gamma_i^* (1 - e^{-d_i l \tau}) + \gamma_i^{**} (1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} < 0 \quad (i = 1, 2),$$
(17)

then the prey-extinction boundary periodic solution $(0, \widetilde{y_1(t)}, 0, \widetilde{y_2(t)})$ of (3) is globally asymptotically stable, where y_i^* (i = 1, 2) and y_i^{**} (i = 1, 2) are defined by (10) and (15).

Proof First, we prove the local stability of the prey-extinction boundary periodic solution $(0, \widetilde{y_1(t)}, 0, \widetilde{y_2(t)})$ of (3). Defining $x_1(t) = x_1(t)$, $y_{11}(t) = y_1(t) - \widetilde{y_1(t)}$, $x_2(t) = x_2(t)$, $y_{12}(t) = y_2(t) - \widetilde{y_2(t)}$, we have the following linearly similar system for (3), which has one periodic solution $(0, \widetilde{y_1(t)}, 0, \widetilde{y_2(t)})$:

$$\begin{pmatrix} \frac{dx_1(t)}{dt} \\ \frac{dy_{11}(t)}{dt} \\ \frac{dx_2(t)}{dt} \\ \frac{dy_{12}(t)}{dt} \end{pmatrix} = \begin{pmatrix} a_1 - \frac{\beta_1 y_1(t)}{\sigma_1} & 0 & 0 & 0 \\ \frac{k_1 \beta_1 y_1(t)}{\sigma_1} & -d_1 & 0 & 0 \\ 0 & 0 & a_2 - \frac{\beta_2 y_2(t)}{\sigma_2} & 0 \\ 0 & 0 & \frac{k_2 \beta_2 y_2(t)}{\sigma_2} & -d_2 \end{pmatrix} \begin{pmatrix} x_1(t) \\ y_{11}(t) \\ x_2(t) \\ y_{12}(t) \end{pmatrix}.$$

It is easy to obtain the fundamental matrix

$$\Phi(t) = \begin{pmatrix} \exp\left[\int_0^t (a_1 - \frac{\beta_1 y_1(s)}{\sigma_1}) ds\right] & 0 & 0 & 0 \\ \exp\left[\int_0^t \frac{k_1 \beta_1 y_1(s)}{\sigma_1} ds\right] & \exp(-d_1 t) & 0 & 0 \\ 0 & 0 & \exp\left[\int_0^t (a_2 - \frac{\beta_2 y_2(s)}{\sigma_2}) ds\right] & 0 \\ 0 & 0 & \exp\left[\int_0^t \frac{k_2 \beta_2 y_2(s)}{\sigma_2} ds\right] & \exp(-d_2 t) \end{pmatrix}.$$

The linearization of the fifth, sixth, seventh, and eighth equations of (3) is

$$\begin{pmatrix} x_1((n+l)\tau^+) \\ y_{11}((n+l)\tau^+) \\ x_2((n+l)\tau^+) \\ y_{12}((n+l)\tau^+) \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1-D & 0 & D \\ 0 & 0 & 1 & 0 \\ 0 & D & 0 & 1-D \end{pmatrix} \begin{pmatrix} x_1((n+l)\tau) \\ y_{11}((n+l)\tau) \\ x_2((n+l)\tau) \\ y_{12}((n+l)\tau) \end{pmatrix}.$$

The linearization of the ninth, tenth, eleventh, and twelfth equations of (3) is

$$\begin{pmatrix} x_1((n+1)\tau^+) \\ y_{11}((n+1)\tau^+) \\ x_2((n+1)\tau^+) \\ y_{12}((n+1)\tau^+) \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1((n+1)\tau) \\ y_{11}((n+1)\tau) \\ x_2((n+1)\tau) \\ y_{12}((n+1)\tau) \end{pmatrix}.$$

The stability of the periodic solution $(0, \widetilde{y_1(t)}, 0, \widetilde{y_2(t)})$ is determined by the eigenvalues of

$$M = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 - D & 0 & D \\ D & 0 & 1 & 0 \\ 0 & D & 0 & 1 - D \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \Phi(\tau),$$

which are

$$\begin{split} \lambda_1 &= \exp\left[\int_0^\tau \left(a_1 - \frac{\beta_1 \tilde{y_1(s)}}{\sigma_1}\right) ds\right],\\ \lambda_2 &= \left|\frac{(1-D)(K_1 + K_3) + \sqrt{(1-D)^2(K_1 + K_3)^2 - 4(1-2D)K_1K_3}}{2}\right|\\ &\leq \left|\frac{(1-D)(K_1 + K_3) + \sqrt{(1+D)^2(K_1 + K_3)^2}}{2}\right|\\ &\leq \left|\frac{(K_1 + K_3)}{2}\right| < 1,\\ \lambda_3 &= \exp\left[\int_0^\tau \left(a_2 - \frac{\beta_2 \tilde{y_2(s)}}{\sigma_2}\right) ds\right], \end{split}$$

and

$$\begin{aligned} \lambda_4 &= \left| \frac{(1-D)(K_1+K_3) - \sqrt{(1-D)^2(K_1+K_3)^2 - 4(1-2D)K_1K_3}}{2} \right| \\ &\leq \left| \frac{(1-D)(K_1+K_3) - \sqrt{(1-D)^2(K_1-K_3)^2}}{2} \right| \\ &= \left| \frac{(1-D)(K_1+K_3) - (1-D)|K_1-K_3|}{2} \right| \\ &\leq (1-D)\max\{K_1,K_3\} < 1, \end{aligned}$$

where $K_1 = e^{-d_1\tau} < 1$, $K_3 = e^{-d_2\tau} < 1$, and condition (16) holds. According to conditions (16), (17), and the Floquet theory [21], if

$$\exp\left[\int_0^\tau \left(a_i - \frac{\beta_i \widetilde{y_i(s)}}{\sigma_i}\right) ds\right] < 1 \quad (i = 1, 2),$$

then

$$\lambda_1 < 1$$

and

$$\lambda_3 < 1$$
,

and thus the prey-extinction boundary periodic solution $(0, \widetilde{y_1(t)}, 0, \widetilde{y_2(t)})$ of (3) is locally stable.

In the following, we will prove the global attraction. By condition (17) we can choose $\varepsilon > 0$ such that

$$\rho_i = \exp\left[\int_0^\tau \left(a_i - \frac{\beta_i(\widetilde{y_i(s)} - \varepsilon)}{\sigma_i}\right) ds\right] < 1 \quad (i = 1, 2).$$

From the second and fourth equations of (3) we notice that $\frac{dy_i(t)}{dt} \ge -d_iy_i(t)$ (*i* = 1, 2). Then, we consider following impulsive comparative differential equation:

$$\begin{cases} \frac{dy_{21}(t)}{dt} = -d_1y_{21}(t), \\ \frac{dy_{22}(t)}{dt} = -d_2y_{22}(t), \end{cases} \quad t \neq (n+l)\tau, t \neq (n+1)\tau, \\ \Delta y_{21}(t) = D(y_{22}(t) - y_{21}(t)), \\ \Delta y_{22}(t) = D(y_{21}(t) - y_{22}(t)), \end{cases} \quad t = (n+l)\tau,$$

$$\begin{cases} \Delta y_{21}(t) = \mu_1, \\ \Delta y_{22}(t) = \mu_2, \end{cases} \quad t = (n+1)\tau. \end{cases}$$
(18)

From Lemma 3.6 and the comparison theorem of impulsive equation (see Theorem 3.1.1 in [21]) we have $y_1(t) \ge y_{21}(t)$, $y_2(t) \ge y_{22}(t)$, and $y_{21}(t) \rightarrow \widetilde{y_1(t)}$, $y_{22}(t) \rightarrow \widetilde{y_2(t)}$ as $t \rightarrow \infty$. Then

$$\begin{cases} y_1(t) \ge y_{21}(t) \ge \widetilde{y_1(t)} - \varepsilon, \\ y_2(t) \ge y_{22}(t) \ge \widetilde{y_2(t)} - \varepsilon \end{cases}$$
(19)

for *t* large enough. For convenience, we may assume that (19) holds for all $t \ge 0$. From (3) and (19) we get

$$\frac{dx_i(t)}{dt} \le \left[a_i - \frac{\beta_i(\widetilde{y_i(t)} - \varepsilon)}{\sigma_i}\right] x_i(t) \quad (i = 1, 2).$$
(20)

So $x_i((n+1)\tau) \le x_i(n\tau^+) \exp\left[\int_{n\tau}^{(n+1)\tau} (a_i - \frac{\beta_i(\widetilde{y_i(s)}-\varepsilon)}{\sigma_i}) ds\right]$ (i = 1, 2). Hence, $x_i(n\tau) \le x_i(0^+)\rho_i^n$ (i = 1, 2) and $x_i(n\tau) \to 0$ (i = 1, 2) as $n \to \infty$; therefore, $x_i(t) \to 0$ (i = 1, 2) as $t \to \infty$.

Next, we will prove that $y_i(t) \to \widetilde{y_i(t)}$ (i = 1, 2) as $t \to \infty$. For $\varepsilon_1 > 0$, there must exist $t_0 > 0$ such that $0 < x_i(t) < \varepsilon_1$ (i = 1, 2) for all $t \ge t_0$. Without loss of generality, we may assume that $0 < x_i(t) < \varepsilon_1$ for all $t \ge 0$. For system (3), we have

$$-d_i y_i(t) \le \frac{dy_i(t)}{dt} \le -\left(d_i - \frac{k_i \beta_i \varepsilon_1}{\sigma_i + \varepsilon_1}\right) y_i(t) \quad (i = 1, 2),$$
(21)

and then we have $y_{21}(t) \le y_1(t) \le y_{31}(t)$, $y_{22}(t) \le y_2(t) \le y_{32}(t)$, and $y_{21}(t) \to y_1(t)$, $y_{22}(t) \to y_2(t)$, $y_{31}(t) \to y_{31}(t)$, $y_{32}(t) \to y_{32}(t)$ as $t \to \infty$, where $(y_{21}(t), y_{22}(t))$ and $(y_{31}(t), y_{32}(t))$ are the solutions of (18) and

$$\begin{cases} \frac{dy_{31}(t)}{dt} = -(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})y_{31}(t),\\ \frac{dy_{32}(t)}{dt} = -(d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})y_{32}(t), \end{cases} \quad t \neq (n+1)\tau, \\ \Delta y_{31}(t) = D(y_{32}(t) - y_{31}(t)),\\ \Delta y_{32}(t) = D(y_{31}(t) - y_{32}(t)), \end{cases} \quad t = (n+1)\tau,$$

$$\Delta y_{31}(t) = \mu_1, \\ \Delta y_{32}(t) = \mu_2, \end{cases} \quad t = (n+1)\tau,$$
(22)

respectively,

$$\begin{split} \widetilde{y_{31}(t)} &= \begin{cases} \widetilde{y_{31}^*e^{-(d_1 - \frac{k_1\beta_1\epsilon_1}{\sigma_1 + \epsilon_1})(t - n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{31}^{**}e^{-(d_1 - \frac{k_1\beta_1\epsilon_1}{\sigma_1 + \epsilon_1})(t - (n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \\ \widetilde{y_{32}(t)} &= \begin{cases} y_{32}^*e^{-(d_2 - \frac{k_2\beta_2\epsilon_1}{\sigma_2 + \epsilon_1})(t - n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{32}^{**}e^{-(d_2 - \frac{k_2\beta_2\epsilon_1}{\sigma_2 + \epsilon_1})(t - (n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \end{cases}$$
(23)

where y_{31}^* and y_{32}^* are determined as

$$\begin{cases} y_{31}^{*} = \frac{\mu_2 B_{31} + \mu_1 (1 - B_{32})}{(1 - A_{31})(1 - B_{32}) - A_{32} B_{31}} > 0, \\ y_{32}^{*} = \frac{\mu_1 A_{32} + \mu_2 (1 - A_{31})}{(1 - A_{31})(1 - B_{32}) - A_{32} B_{31}} > 0, \end{cases}$$
(24)

and y_{31}^{**} and y_{32}^{**} are defined as

$$\begin{cases} y_{31}^{**} = (1-D)e^{-(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})l\tau} y_{31}^* + De^{-(d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})l\tau} y_{32}^*, \\ y_{32}^{**} = De^{-(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})l\tau} y_{31}^* + (1-D)e^{-(d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})l\tau} y_{32}^*, \end{cases}$$
(25)

where

$$\begin{split} A_{31} &= (1-D)e^{-(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})\tau} < 1, \\ B_{31} &= De^{-[(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})(1-l) + (d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})l]\tau} < 1, \\ A_{32} &= De^{-[(d_1 - \frac{k_1\beta_1\varepsilon_1}{\sigma_1 + \varepsilon_1})l + (d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})(1-l)]\tau} < 1, \\ B_{32} &= (1-D)e^{-(d_2 - \frac{k_2\beta_2\varepsilon_1}{\sigma_2 + \varepsilon_1})\tau} < 1. \end{split}$$

For any $\varepsilon_2 > 0$, there exists $t_1, t > t_1$, such that

$$\widetilde{y_{21}(t)} - \varepsilon_2 < y_1(t) < \widetilde{y_{31}(t)} + \varepsilon_2$$

and

$$\widetilde{y_{22}(t)} - \varepsilon_2 < y_2(t) < \widetilde{y_{32}(t)} + \varepsilon_2.$$

Letting $\varepsilon_1 \rightarrow 0$, we have

$$\widetilde{y_1(t)} - \varepsilon_2 < y_1(t) < \widetilde{y_1(t)} + \varepsilon_2$$

and

$$\widetilde{y_2(t)} - \varepsilon_2 < y_2(t) < \widetilde{y_2(t)} + \varepsilon_2$$

for *t* large enough, which implies $y_1(t) \to y_1(t)$ and $y_2(t) \to y_2(t)$ as $t \to \infty$. This completes the proof.

The next work is to investigate the permanence of system (3).

Definition 4.2 System (3) is said to be permanent if there are constants m, M > 0 (independent of initial value) and a finite time T_0 such that for all solutions $(x_1(t), y_1(t), x_2(t), y_2(t))$ with any initial values $x_1(0^+) > 0$, $y_1(0^+) > 0$, $x_2(0^+) > 0$, $y_2(0^+) > 0$, we have $m \le x_1(t) \le M$, $m \le y_1(t) \le M$, $m \le x_2(t) \le M$, $m \le y_2(t) \le M$ for all $t \ge T_0$. Here T_0 may depend on the initial values $(x_1(0^+), y_1(0^+), x_2(0^+), y_2(0^+))$.

Theorem 4.3 If

$$\min_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [y_i^* (1 - e^{-d_i l \tau}) + y_i^{**} (1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} > 0 \quad (i = 1, 2),$$
(26)

then system (3) is permanent, where y_i^* (i = 1, 2) and y_i^{**} (i = 1, 2) are defined by (10) and (15), respectively.

Proof Suppose $(x_1(t), y_1(t), x_2(t), y_2(t))$ is a solution of (3) with $x_1(0) > 0, y_1(0) > 0, x_2(0) > 0$, $y_2(0) > 0$. By Lemma 3.4 there exists a constant M > 0 such that $x_1(t) \le M$, $y_1(t) \le M$, $x_2(t) \le M$, for t large enough. From (3) and Theorem 4.1 we have $y_i(t) > y_i(t) - \varepsilon_2 > y_i^* e^{-d_i l\tau} + y_i^{**} e^{-d_i (1-l)\tau} \triangleq m_i \ (i = 1, 2)$ for ε_2 small enough. So we only need to find $m_3 > 0$ and ε_3 such that $x_i(t) > m_3$ for t large enough. Otherwise, we can select $m_4 > 0$ small enough satisfying $m_4 < \frac{\sigma_i d_i}{k_i \beta_i - d_i} \ (d_i < k_i \beta_i)$ and prove that $x_i(t) < m_4$ cannot hold for $t \ge 0$. Suppose the contrary. By condition (26), choosing ε_3 small enough, we can obtain

$$\delta_{i} = a_{i}\tau - \frac{\beta_{i}[y_{4i}^{*}(1 - e^{-(d_{i} - \frac{k_{i}\beta_{i}m_{4}}{\sigma_{i} + m_{4}})l\tau}) + y_{4i}^{**}(1 - e^{-(d_{i} - \frac{k_{i}\beta_{i}m_{4}}{\sigma_{i} + m_{4}})(1 - l)\tau})]}{\sigma_{i}(d_{i} - \frac{k_{i}\beta_{i}m_{4}}{\sigma_{i} + m_{4}})} - \frac{\beta_{i}\varepsilon_{3}}{\sigma_{i}}\tau > 0$$

with y_{4i}^{*} (*i* = 1, 2) and y_{4i}^{**} (*i* = 1, 2) are defined by (30) and (31). Then,

$$\begin{cases} \frac{dy_{1}(t)}{dt} < -(d_{1} - \frac{k_{1}\beta_{1}m_{4}}{\sigma_{1}+m_{4}})y_{1}(t),\\ \frac{dy_{2}(t)}{dt} < -(d_{2} - \frac{k_{2}\beta_{2}m_{4}}{\sigma_{2}+m_{4}})y_{2}(t), \end{cases} \quad t \neq (n+l)\tau, t \neq (n+1)\tau, \\ \Delta y_{1}(t) = D(y_{2}(t) - y_{1}(t)),\\ \Delta y_{2}(t) = D(y_{1}(t) - y_{2}(t)), \end{cases} \quad t = (n+l)\tau, \tag{27}$$

$$\begin{cases} \Delta y_{1}(t) = \mu_{1},\\ \Delta y_{2}(t) = \mu_{2}, \end{cases} \quad t = (n+1)\tau. \end{cases}$$

By Lemma 3.6 we have $y_1(t) \le y_{41}(t), y_2(t) \le y_{42}(t)$ and $y_{41}(t) \to \overline{y_{41}(t)}, y_{42}(t) \to \overline{y_{42}(t)}, t \to \infty$, where $(y_{41}(t), y_{42}(t))$ is the solution of

$$\begin{cases} \frac{dy_{41}(t)}{dt} = -(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})y_{41}(t),\\ \frac{dy_{42}(t)}{dt} = -(d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})y_{42}(t), \end{cases} \quad t \neq (n+1)\tau, \\ \Delta y_{41}(t) = D(y_{42}(t) - y_{41}(t)),\\ \Delta y_{42}(t) = D(y_{41}(t) - y_{42}(t)), \end{cases} \quad t = (n+1)\tau,$$

$$\Delta y_{41}(t) = \mu_1,\\ \Delta y_{42}(t) = \mu_2, \end{cases} \quad t = (n+1)\tau,$$
(28)

with

$$\begin{cases} \overline{y_{41}(t)} = \begin{cases} y_{41}^* e^{-(d_1 - \frac{k_1 \beta_1 m_4}{\sigma_1 + m_4})(t - n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{41}^{**} e^{-(d_1 - \frac{k_1 \beta_1 m_4}{\sigma_1 + m_4})(t - (n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \\ \hline y_{42}(t) = \begin{cases} y_{42}^* e^{-(d_2 - \frac{k_2 \beta_2 m_4}{\sigma_2 + m_4})(t - n\tau)}, & t \in (n\tau, (n+l)\tau], \\ y_{42}^{**} e^{-(d_2 - \frac{k_2 \beta_2 m_4}{\sigma_2 + m_4})(t - (n+l)\tau)}, & t \in ((n+l)\tau, (n+1)\tau], \end{cases}$$
(29)

where y_{41}^* and y_{42}^* are determined as

$$\begin{cases} y_{41}^{*} = \frac{\mu_{2}B_{41} + \mu_{1}(1 - B_{42})}{(1 - A_{41})(1 - B_{42}) - A_{42}B_{41}} > 0, \\ y_{42}^{*} = \frac{\mu_{1}A_{42} + \mu_{2}(1 - A_{41})}{(1 - A_{41})(1 - B_{42}) - A_{42}B_{41}} > 0, \end{cases}$$
(30)

and y_{41}^{**} , y_{42}^{**} are defined as

$$\begin{cases} y_{41}^{**} = (1-D)e^{-(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})l\tau} y_{41}^* + De^{-(d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})l\tau} y_{42}^*, \\ y_{42}^{**} = De^{-(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})l\tau} y_{41}^* + (1-D)e^{-(d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})l\tau} y_{42}^*, \end{cases}$$
(31)

where

$$\begin{split} A_{41} &= (1-D)e^{-(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})\tau} < 1, \\ B_{41} &= De^{-[(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})(1-l) + (d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})l]\tau} < 1, \\ A_{42} &= De^{-[(d_1 - \frac{k_1\beta_1m_4}{\sigma_1 + m_4})l + (d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})(1-l)]\tau} < 1, \\ B_{42} &= (1-D)e^{-(d_2 - \frac{k_2\beta_2m_4}{\sigma_2 + m_4})\tau} < 1. \end{split}$$

Therefore, there exist $T_1 > 0$ and $\varepsilon_3 > 0$ such that

$$y_1(t) \le y_{41}(t) \le \overline{y_{41}(t)} + \varepsilon_3$$

and

$$y_2(t) \le y_{42}(t) \le \overline{y_{42}(t)} + \varepsilon_3.$$

Then,

$$\frac{dx_i(t)}{dt} \ge \left[a_i - \frac{\beta_i(\overline{y_{4i}(t)} + \varepsilon_3)}{\sigma_i}\right] x_i(t) \quad (i = 1, 2)$$
(32)

for $t \ge T_1$. Let $N_1 \in N$ and $N_1\tau > T_1$. Integrating (32) on $(n\tau, (n+1)\tau)$, $n \ge N_1$, we have

$$\begin{aligned} x_i\big((n+1)\tau\big) &\geq x_i\big(n\tau^+\big)\exp\left(\int_{n\tau}^{(n+1)\tau} \left[a_i - \frac{\beta_i(\overline{y_{4i}(t)} + \varepsilon_3)}{\sigma_i}\right]dt\right) \\ &= x_i(n\tau)e^{\delta_i} \quad (i=1,2). \end{aligned}$$

Then, $x_i((N_1 + k)\tau) \ge x_i(N_1\tau^+)e^{k\delta_i} \to \infty$ as $k \to \infty$, which is a contradiction to the boundedness of $x_i(t)$ (i = 1, 2). Hence, there exists $t_1 > 0$ such that $x_i(t) \ge m_3$ (i = 1, 2). This completes the proof.

5 Simulation analysis and discussion

In this paper, we establish a predator-prey model with impulsive diffusion and releasing on predator population. This predator-prey model for two regions, which are connected by diffusion of predator population, portrays the evolvement of population. We prove that all solutions of the investigated system are uniformly ultimately bounded. By Theorem 4.1, if $D < \frac{1}{2}$ and

$$\max_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [y_i^* (1 - e^{-d_i l \tau}) + y_i^{**} (1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} < 0 \quad (i = 1, 2),$$

then the prey-extinction boundary periodic solution $(0, y_1(t), 0, y_2(t))$ of system (3) is globally asymptotically stable. By Theorem 4.3, if

$$\min_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [y_i^*(1 - e^{-d_i l \tau}) + y_i^{**}(1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} > 0 \quad (i = 1, 2),$$

then system (3) is permanent.





5.1 The dynamical behaviors influenced by parameter D

Let $x_1(0) = 0.5$, $y_1(0) = 0.5$, $x_2(0) = 0.5$, $y_2(0) = 0.5$, $a_1 = 0.1$, $b_1 = 0.2$, $a_2 = 0.1$, $b_2 = 0.2$, $\beta_1 = 0.5$, $\beta_2 = 5$, $k_1 = 0.5$, $k_2 = 5$, $\mu_1 = 0.5$, $\mu_2 = 0.3$, $d_1 = 0.3$, $d_2 = 0.3$, $\sigma_1 = 3.5$, $\sigma_2 = 3.5$, $\tau = 1$, l = 0.25, D = 0.1. Then conditions (16) and (17) are obviously satisfied, and thus the preyextinction periodic solution of system (3) is globally asymptotically stable (see Figure 1). Also assume that $x_1(0) = 0.5$, $y_1(0) = 0.5$, $x_2(0) = 0.5$, $y_2(0) = 0.5$, $a_1 = 0.1$, $b_1 = 0.2$, $a_2 = 0.1$, $b_2 = 0.2$, $\beta_1 = 0.5$, $\beta_2 = 5$, $k_1 = 0.5$, $k_2 = 5$, $\mu_1 = 0.5$, $\mu_2 = 0.3$, $d_1 = 0.3$, $d_2 = 0.3$, $\sigma_1 = 3.5$, $\sigma_2 = 3.5$, $\tau = 1$, l = 0.25, D = 0.95. Then condition (26) is obviously satisfied, and system (3) is permanent (see Figure 2). From (17) and (26) we can calculate that there exists one threshold D^* , which satisfies

$$\max_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [y_i^* (1 - e^{-d_i l \tau}) + y_i^{**} (1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} < 0 \quad (i = 1, 2)$$

or

$$\min_{i=1,2} \left\{ a_i \tau - \frac{\beta_i [y_i^* (1 - e^{-d_i l \tau}) + y_i^{**} (1 - e^{-d_i (1 - l) \tau})]}{\sigma_i d_i} \right\} > 0 \quad (i = 1, 2).$$

If $D > D^*$, then the prey population will go to extinction; if $D < D^*$, then the population will be permanent.







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5.2 The dynamical behaviors influenced by parameters μ_1 and μ_2

In this subsection, we always assume that $\mu = \mu_1 = \mu_2$. Assume that $x_1(0) = 0.5$, $y_1(0) = 0.5$, $x_2(0) = 0.5$, $y_2(0) = 0.5$, $a_1 = 0.1$, $b_1 = 0.2$, $a_2 = 0.1$, $b_2 = 0.2$, $\beta_1 = 0.5$, $\beta_2 = 5$, $k_1 = 0.5$, $k_2 = 5$, $\mu_1 = 0.4$, $\mu_2 = 0.4$, $d_1 = 0.4$, $d_2 = 0.3$, $\sigma_1 = 3.5$, $\sigma_2 = 3.5$, $\tau = 1$, l = 0.25, D = 0.2. Then conditions (16) and (17) are obviously satisfied, and the prey-extinction periodic solution of system (3) is globally asymptotically stable (see Figure 3). Also, assume that $x_1(0) = 0.5$, $y_1(0) = 0.5$, $x_2(0) = 0.5$, $y_2(0) = 0.5$, $a_1 = 0.1$, $b_1 = 0.2$, $a_2 = 0.1$, $b_2 = 0.2$, $\beta_1 = 0.5$, $\beta_2 = 5$, $k_1 = 0.5$, $k_2 = 5$, $\mu_1 = 0.3$, $\mu_2 = 0.3$, $d_1 = 0.3$, $d_2 = 0.3$, $\sigma_1 = 3.5$, $\sigma_2 = 3.5$, $\tau = 1$, l = 0.25, D = 0.2. Then condition (26) is obviously satisfied, and system (3) is permanent (see Figure 4). We can calculate that there exists at least one threshold μ^* such that if $\mu > \mu^*$, then the prey population will go to extinction, and if $\mu < \mu^*$, then the population will be permanent.

From the simulations we discover that the increasing diffusive rate of predator population will count against the pest management. We conclude that the impulsive diffusion and releasing predator provide reliable tactic basis for pest management.

Competing interests

The authors declare that they have no competing interests.

Authors' contributions

AZ carried out the main part of this article, PS corrected the manuscript, JJ brought forward many suggestions on this article. All authors have read and approved the final manuscript.

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