



*Research article*

## **Almost periodic solutions for a SVIR epidemic model with relapse**

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**Abstract:** This paper is devoted to a nonautonomous SVIR epidemic model with relapse, that is, the recurrence rate is considered in the model. The permanent of the system is proved, and the result on the existence and uniqueness of globally attractive almost periodic solution of this system is obtained by constructing a suitable Lyapunov function. Some analysis for the necessity of considering the recurrence rate in the model is also presented. Moreover, some examples and numerical simulations are given to show the feasibility of our main results. Through numerical simulation, we have obtained the influence of vaccination rate and recurrence rate on the spread of the disease. The conclusion is that in order to control the epidemic of infectious diseases, we should increase the vaccination rate while reducing the recurrence rate of the disease.

**Keywords:** epidemic model; persistence; almost periodic solution

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### **1. Introduction**

As early as the 2nd century AD, the Antonine plague was prevalent in the Roman Empire, causing a rapid population decline and economic deterioration, allowing invaders to take advantage of it and leading to the collapse of the Roman Empire. Between 1519–1530 AD, an epidemic of measles and other infectious diseases reduced Mexico's Indian population from 30 million to 3 million. Since the outbreak of COVID-19, the number of infected people worldwide has reached more than 140 million, many countries have prevented the spread of the virus by reducing people's contact, but these measures have had a huge impact on the economy. These shows that infectious diseases have always been the primary factor affecting human death. Therefore, it is necessary to study the transmission mechanism and the dynamic behavior of epidemics.

Inspired by the Kermack and Mckendricks' original epidemic model [1], the study of mathematical epidemiology has grown rapidly, with a large variety of models having been formulated and applied to infectious diseases. For example, we can see SIS (susceptible-infectious-susceptible) models in [2], the authors considered the global dynamics of the SIS model with delays denoting an incubation time.

By constructing a Lyapunov function, they prove stability of a disease-free equilibrium  $E_0$  under a condition different from other paper. There are some other SIS models, we can refer to [3, 4] and the references cited therein. Kuniya etc. [5] formulated a SIR (susceptible-infectious-recovered) model with nonlocal diffusion. They prove the global asymptotic stability of the disease-free equilibrium when the basic reproduction number  $R_0 < 1$  and also prove the uniform persistence of the system when  $R_0 > 1$  by using the persistent theory for dynamical systems. In 2020, Naik et al. [6] established a new SIR model with Crowley-Martin type functional response and Holling type II treatment rate. The authors proved that the model has a disease-free equilibrium point and an endemic equilibrium point, and studied the existence and stability of the equilibrium point by using the La-Salle invariance principle and the Lyapunov function. The author also established a similar model in the same year, please refer to [7]. There are some other SIR models, we can refer to [8] and the references cited therein. In addition to these, the application of fractional-order models in infectious diseases has become more and more widespread. Naik et al. [9] established a nonlinear fractional infectious disease model for HIV transmission. In the analysis process, the authors introduced Caputo-type fractional derivatives, applied the generalized Adams-Bashforth-Moulton method to find the numerical solution of the model. And by using the fractional Routh-Hurwitz stability criterion and the fractional La-Salle invariance principle, the equilibrium state of the model is determined and its stability is analyzed. Similar models can also refer to [10–12] and the references cited therein. In addition, some scholars have considered treatment items in the model, such as the SEIR (susceptible-exposed-infectious-recovered) model in [13, 14]. In addition to these, there are other models that take into account more factors, as in [15–19]. From the work of S.A. Boone, et al. [20] and A. Gabbuti, et al. [21], we know that vaccination is the most effective measure to control the spread of hepatitis B due to prevention and management of viral disease heavily relies upon vaccines and antiviral medications. In 2014, Xu et al. [22] studied vaccination decisions based on game theory and its impact on the spread of infectious diseases. In 2016, they studied the importance of vaccination for sexually transmitted diseases. It is concluded that with the establishment of herd immunity after vaccination, the mortality rate caused by infectious diseases will decrease [23]. In 2021, scholars have also analyzed the impact of vaccination on the spread of COVID-19 [24, 25]. There are many similar models, see [26–28]. Considering a continuous vaccination strategy, Liu et al. [28] formulated the following system of ordinary differential equations:

$$\begin{cases} \frac{dS(t)}{dt} = \mu - \beta S(t)I(t) - (\mu + \alpha)S(t), \\ \frac{dV(t)}{dt} = \alpha S(t) - \beta_1 V(t)I(t) - (\mu + \gamma_1)V(t), \\ \frac{dI(t)}{dt} = \beta S(t)I(t) + \beta_1 V(t)I(t) - \gamma I(t) - \mu I(t), \\ \frac{dR(t)}{dt} = \gamma_1 V(t) + \gamma I(t) - \mu R(t), \end{cases} \quad (1.1)$$

where  $S(t)$ ,  $V(t)$ ,  $I(t)$  and  $R(t)$  denote the susceptible, vaccinated, infectious and recovered populations as time  $t$ , respectively. The parameters  $\mu$ ,  $\beta$ ,  $\alpha$ ,  $\beta_1$ ,  $\gamma_1$ ,  $\gamma$  are all positive constants, here,  $\mu$  is the recruitment rate and natural death rate,  $\beta$  is the rate of disease transmission between susceptible and infectious individuals,  $\beta_1$  is the rate of disease transmission between vaccinated and infected individuals,  $\gamma$  is the recovery rate,  $\alpha$  is the vaccination rate,  $\gamma_1$  is the rate at which a vaccinated individual obtains immunity. It was shown that the global dynamics of model (1.1) is completely determined by the basic

reproduction number  $R_0$ . That is, if  $R_0 < 1$ , the disease-free equilibrium is globally asymptotically stable, while if  $R_0 > 1$ , a positive endemic equilibrium exists and it is globally asymptotically stable. It was observed in [28] that vaccination has an effect of decreasing the basic reproduction number.

We all know that some infectious diseases, such as measles and mumps, will be immune to life after being caught once, but not all infectious diseases will remain immune to death after being acquired once. Neglecting to monitor those who have recovered may lead to a resurgence of the disease. Therefore, it is necessary to consider the recurrence rate of the disease in the model.

Many previous models such as model (1.1) have constant coefficients. Assuming that the coefficient is independent of the environment is not consistent with the actual situation. Parameters (death rate, birth rate, etc.) are subject to changes in season, weather, food supply, etc. Incidence rates of many infectious diseases, such as measles, chicken-pox, rubella, diphtheria and influenza, are periodic or almost periodic in nature [29], therefore, nonautonomous systems are more realistic to reflect actual problems. Recently, researchers have worked on the nonautonomous epidemic dynamical systems with almost periodic parameters [30–37], the concept of almost periodicity was introduced by Bohr [38].

Motivated by the above work, in this paper, we consider the following nonautonomous SVIR epidemic model with relapse:

$$\begin{cases} \frac{dS(t)}{dt} = \mu(t) - \beta(t)S(t)I(t) - (\mu(t) + \alpha(t))S(t), \\ \frac{dV(t)}{dt} = \alpha(t)S(t) - \beta_1(t)V(t)I(t) - (\mu(t) + \gamma_1(t))V(t), \\ \frac{dI(t)}{dt} = \beta(t)S(t)I(t) + \beta_1(t)V(t)I(t) - \gamma(t)I(t) - \mu(t)I(t) + k(t)R(t), \\ \frac{dR(t)}{dt} = \gamma_1(t)V(t) + \gamma(t)I(t) - \mu(t)R(t) - k(t)R(t), \end{cases} \quad (1.2)$$

for  $t \in \mathbb{R}^+ = [0, +\infty)$ , with initial conditions:

$$S(0) > 0, V(0) > 0, I(0) > 0, R(0) > 0,$$

where  $\mu(t), \alpha(t), \beta(t), \beta_1(t), \gamma(t), \gamma_1(t), k(t)$  are positive almost periodic functions for  $t \in \mathbb{R}$ , the ecological meaning of the almost functions are described in Table 1.

**Table 1.** The ecological meaning of the functions.

Almost functions	Ecological meaning
$\mu(t)$	Recruitment rate and natural death rate of the population
$\alpha(t)$	Vaccination rate of the population
$\beta(t)$	The rate of disease transmission between $S$ and $I$
$\beta_1(t)$	The rate of disease transmission between $V$ and $I$
$\gamma(t)$	The recovery rate
$\gamma_1(t)$	The rate at which a vaccinated individual obtains immunity
$k(t)$	The recurrence rate

The paper is organized as follows: In section 2, some definitions and lemmas is presented. Section 3 is devoted to the permanence of the system (1.2). Section 4 is mainly to get the uniqueness and global

attractivity of almost periodic solution of the system (1.2). Section 5 mainly analyzes the necessity of adding the recurrence rate  $k$  in this model and discusses the relationship between the recurrence rate  $k$  and the basic reproduction number  $R_0$ . Some numerical simulations in Section 6 and discussions in Section 7 are given to illustrate our analytical result.

## 2. Preliminaries

In this section, some definitions and lemmas will be presented.

**Definition 1.** [38,39] A continuous function  $f : \mathbb{R} \rightarrow \mathbb{R}$  is said to be almost periodic on  $\mathbb{R}$  if for any  $\epsilon > 0$ , the set

$$T(f, \epsilon) \equiv \{\tau \in \mathbb{R} : |f(t + \tau) - f(t)| < \epsilon, t \in \mathbb{R}\}$$

is relatively dense in  $\mathbb{R}$ . i.e., for any  $\epsilon > 0$ , it is possible to find a real number  $l = l(\epsilon) > 0$  with the property that for any interval  $L$  with length  $l(\epsilon)$  such that  $L \cap T(f, \epsilon) \neq \emptyset$ .

**Definition 2.** [38,39] A continuous function  $f : \mathbb{R}^+ \rightarrow \mathbb{R}$  is said to be asymptotically almost periodic function if there exist an almost periodic function  $h(t)$  and a continuous function  $\varphi(t)$  defined on  $\mathbb{R}^+$  with  $\lim_{t \rightarrow \infty} \varphi(t) = 0$  such that

$$f(t) = h(t) + \varphi(t).$$

Similar to almost periodic functions, asymptotic almost periodic functions also have several equivalent definitions:

**Proposition 1.** [38,39] Let  $f : \mathbb{R}^+ \rightarrow \mathbb{R}$  be a continuous function,  $f$  is asymptotically almost periodic if and only if for any  $\epsilon > 0$ ,  $\mathcal{T}(\epsilon) \geq 0$ , the set

$$T^+(f, \epsilon) \equiv \{\tau \in \mathbb{R}^+ : |f(t + \tau) - f(t)| < \epsilon, t \geq T(\epsilon), t + \tau \geq \mathcal{T}(\epsilon)\}$$

is relatively dense in  $\mathbb{R}^+$ . i.e., for any  $\epsilon > 0$ , it is possible to find a real number  $l = l(\epsilon) > 0$  with the property that for any interval  $L \subset \mathbb{R}^+$  with length  $l(\epsilon) > 0$  and  $\mathcal{T}(\epsilon) \geq 0$  such that  $L \cap T^+(f, \epsilon) \neq \emptyset$ .

**Lemma 1.** [40] If  $a > 0, b > 0$  and  $\frac{dx(t)}{dt} \geq b - ax$  ( $\frac{dx(t)}{dt} \leq b - ax$ ), where  $t \geq t_0$  and  $x(t_0) > 0$ , we have

$$\liminf_{t \rightarrow \infty} x(t) \geq \frac{b}{a} \quad \left( \limsup_{t \rightarrow \infty} x(t) \leq \frac{b}{a} \right).$$

**Lemma 2.** [41,42] If function  $f$  is nonnegative, integrable and uniformly continuous on  $[0, +\infty]$ , then  $\lim_{t \rightarrow \infty} f(t) = 0$ .

## 3. Permanence

For convenience, we introduce some notations, in the following part of this paper, let  $g : \mathbb{R} \rightarrow \mathbb{R}$  be a continuous bounded function, then  $\bar{g}$  and  $\underline{g}$  will be defined as

$$\bar{g} = \sup_{t \in \mathbb{R}} g(t), \quad \underline{g} = \inf_{t \in \mathbb{R}} g(t).$$

It is well known that an almost periodic function is bounded and uniformly continuous, thus we have

$$\max\{\bar{\mu}, \bar{\alpha}, \bar{\beta}, \bar{\beta}_1, \bar{\gamma}, \bar{\gamma}_1, \bar{k}\} < \infty.$$

Moreover, we always assume that

$$\min\{\underline{\mu}, \underline{\alpha}, \underline{\beta}, \underline{\beta}_1, \underline{\gamma}, \underline{\gamma}_1, \underline{k}\} > 0.$$

**Theorem 1.** *The system (1.2) is permanent, it means that any solution  $(S(t), V(t), I(t), R(t))$  of system (1.2) satisfies*

$$\begin{aligned} m_1 &\leq \liminf_{t \rightarrow \infty} S(t) \leq \limsup_{t \rightarrow \infty} S(t) \leq M_1, \\ m_2 &\leq \liminf_{t \rightarrow \infty} V(t) \leq \limsup_{t \rightarrow \infty} V(t) \leq M_2, \\ m_3 &\leq \liminf_{t \rightarrow \infty} I(t) \leq \limsup_{t \rightarrow \infty} I(t) \leq M_3, \\ m_4 &\leq \liminf_{t \rightarrow \infty} R(t) \leq \limsup_{t \rightarrow \infty} R(t) \leq M_4, \end{aligned} \quad (3.1)$$

where  $M_1 = \frac{\bar{\mu}}{\underline{\mu} + \underline{\alpha}}$ ,  $M_2 = \frac{\bar{\alpha}M_1}{\underline{\mu} + \underline{\gamma}_1}$ ,  $M_3 = \frac{\bar{\mu}}{\underline{\mu}}$ ,  $M_4 = \frac{\bar{\gamma}_1 M_2 + \bar{\gamma} M_3}{\underline{\mu} + \underline{k}}$ ,  $m_1 = \frac{\underline{\mu}}{\bar{\beta}M_2 + \bar{\mu} + \bar{\alpha}}$ ,  $m_2 = \frac{\underline{\alpha}m_1}{\bar{\beta}_1 M_3 + \bar{\mu} + \bar{\gamma}_1}$ ,  $m_3 = \frac{\underline{k}m_4}{\bar{\mu} + \bar{\gamma}}$ ,  $m_4 = \frac{\underline{\gamma}_1 m_2}{\bar{\mu} + \bar{k}}$ .

*Proof.* The first equation of system (1.2) gives that

$$S(t) = S(0)e^{\int_0^t (\frac{\mu(y)}{S(y)} - \beta(y)I(y) - \mu(y) - \alpha(y))dy},$$

where  $S(0) > 0$ . This implies that  $S(t) > 0$  for all  $t > 0$ , similar results also hold for  $V(t)$ ,  $I(t)$  and  $R(t)$ .

From the first equation of system (1.2), we have

$$\begin{aligned} \frac{dS(t)}{dt} &\leq \mu(t) - (\mu(t) + \alpha(t))S(t) \\ &\leq \bar{\mu} - (\underline{\mu} + \underline{\alpha})S(t). \end{aligned} \quad (3.2)$$

By applying Lemma 1 to (3.2), we get

$$\limsup_{t \rightarrow \infty} S(t) \leq \frac{\bar{\mu}}{\underline{\mu} + \underline{\alpha}} \triangleq M_1.$$

Thus, there exists a sufficiently small  $\epsilon_1 > 0$  and  $T_1 > 0$  such that

$$S(t) \leq M_1 + \epsilon_1, \quad t \geq T_1.$$

According to the second equation of system (1.2), we get

$$\begin{aligned} \frac{dV(t)}{dt} &\leq \alpha(t)S(t) - (\mu(t) + \gamma_1(t))V(t) \\ &\leq \bar{\alpha}(M_1 + \epsilon_1) - (\underline{\mu} + \underline{\gamma}_1)V(t). \end{aligned} \quad (3.3)$$

By applying Lemma 1 to (3.3), we get

$$\limsup_{t \rightarrow \infty} V(t) \leq \frac{\bar{\alpha}(M_1 + \epsilon_1)}{\underline{\mu} + \underline{\gamma}_1} \rightarrow \frac{\bar{\alpha}M_1}{\underline{\mu} + \underline{\gamma}_1} \triangleq M_2(\epsilon_1 \rightarrow 0).$$

So as  $\epsilon_1 \rightarrow 0$  there exists a sufficiently small  $\epsilon_2 > 0$  and  $T_2 > T_1$  such that

$$V(t) \leq M_2 + \epsilon_2, t \geq T_2.$$

Let  $N(t) = S(t) + V(t) + I(t) + R(t)$ , then we get

$$\begin{aligned} \frac{dN(t)}{dt} &= \mu(t) - \mu(t)N(t) \\ &\leq \bar{\mu} - \underline{\mu}N(t). \end{aligned} \quad (3.4)$$

By applying Lemma 1 to (3.4), we get

$$\limsup_{t \rightarrow \infty} N(t) \leq \frac{\bar{\mu}}{\underline{\mu}} \triangleq M_3.$$

Then we have

$$\limsup_{t \rightarrow \infty} I(t) \leq M_3.$$

Thus, there exists a sufficiently small  $\epsilon_3 > 0$  and  $T_3 > T_2$  such that

$$I(t) \leq M_3 + \epsilon_3, t \geq T_3.$$

From the last equation of the system (1.2), we get

$$\frac{dR(t)}{dt} \leq \bar{\gamma}_1(M_2 + \epsilon_2) + \bar{\gamma}(M_3 + \epsilon_3) - (\underline{\mu} + \underline{k})R(t). \quad (3.5)$$

By applying Lemma 1 to (3.5), we get

$$\limsup_{t \rightarrow \infty} R(t) \leq \frac{\bar{\gamma}_1(M_2 + \epsilon_2) + \bar{\gamma}(M_3 + \epsilon_3)}{\underline{\mu} + \underline{k}} \rightarrow \frac{\bar{\gamma}_1 M_2 + \bar{\gamma} M_3}{\underline{\mu} + \underline{k}} \triangleq M_4(\epsilon_2, \epsilon_3 \rightarrow 0).$$

So as  $\epsilon_2, \epsilon_3 \rightarrow 0$ , there exists a sufficiently small  $\epsilon_4 > 0$  and  $T_4 > T_3$  such that

$$R(t) \leq M_4 + \epsilon_4, t \geq T_4.$$

Similarly, we can get the following inequation by using the above conclusions,

$$\frac{dS(t)}{dt} \geq \underline{\mu} - (\bar{\beta}(M_2 + \epsilon_2) + \bar{\mu} + \bar{\alpha})S(t). \quad (3.6)$$

By applying Lemma 1 to (3.6), we get

$$\liminf_{t \rightarrow \infty} S(t) \geq \frac{\underline{\mu}}{\bar{\beta}(M_2 + \epsilon_2) + \bar{\mu} + \bar{\alpha}} \rightarrow \frac{\underline{\mu}}{\bar{\beta}M_2 + \bar{\mu} + \bar{\alpha}} \triangleq m_1(\epsilon_2 \rightarrow 0).$$

So as  $\epsilon_2 \rightarrow 0$ , there exists a sufficiently small  $\epsilon_5 > 0$  and  $T_5 > T_4$  such that

$$S(t) \geq m_1 - \epsilon_5, t \geq T_5.$$

According to the second equation of system (1.2), we get

$$\frac{dV(t)}{dt} \geq \underline{\alpha}(m_1 - \epsilon_5) - (\bar{\beta}_1(M_3 + \epsilon_3) + \bar{\mu} + \bar{\gamma}_1)V(t). \quad (3.7)$$

By applying Lemma 1 to (3.7), as  $\epsilon_5 \rightarrow 0$ , we get

$$\liminf_{t \rightarrow \infty} V(t) \geq \frac{\underline{\alpha}(m_1 - \epsilon_5)}{\bar{\beta}_1(M_3 + \epsilon_3) + \bar{\mu} + \bar{\gamma}_1} \rightarrow \frac{\underline{\alpha}m_1}{\bar{\beta}_1M_3 + \bar{\mu} + \bar{\gamma}_1} \triangleq m_2(\epsilon_3, \epsilon_5 \rightarrow 0).$$

So as  $\epsilon_3, \epsilon_5 \rightarrow 0$ , there exists a sufficiently small  $\epsilon_6 > 0$  and  $T_6 > T_5$  such that

$$V(t) \geq m_2 - \epsilon_6, t \geq T_6.$$

From the last equation of the system (1.2), we get

$$\frac{dR(t)}{dt} \geq \underline{\gamma}_1(m_2 - \epsilon_6) - (\bar{\mu} + \bar{k})R(t). \quad (3.8)$$

By applying Lemma 1 to (3.8), we get

$$\liminf_{t \rightarrow \infty} R(t) \geq \frac{\underline{\gamma}_1(m_2 - \epsilon_6)}{\bar{\mu} + \bar{k}} \rightarrow \frac{\underline{\gamma}_1 m_2}{\bar{\mu} + \bar{k}} \triangleq m_4(\epsilon_6 \rightarrow 0).$$

Thus, there exists a sufficiently small  $\epsilon_7 > 0$  and  $T_7 > T_6$  such that

$$R(t) \geq m_4 - \epsilon_7, t \geq T_7.$$

From the third equation of the system (1.2), we get

$$\frac{dI(t)}{dt} \geq \underline{k}(m_4 - \epsilon_7) - (\bar{\mu} + \bar{\gamma})I(t). \quad (3.9)$$

By applying Lemma 1 to (3.9), we get

$$\liminf_{t \rightarrow \infty} I(t) \geq \frac{\underline{k}(m_4 - \epsilon_7)}{\bar{\mu} + \bar{\gamma}} \rightarrow \frac{\underline{k}m_4}{\bar{\mu} + \bar{\gamma}} \triangleq m_3(\epsilon_7 \rightarrow 0).$$

Thus, there exists a sufficiently small  $\epsilon_8 > 0$  and  $T_8 > T_7$  such that

$$I(t) \geq m_3 - \epsilon_8, t \geq T_8.$$

□

#### 4. Almost periodic solution

**Theorem 2.** Suppose that the system satisfies the following conditions:

$$\begin{cases} \underline{\mu} - 2\bar{\beta}M_3 > 0, \\ \underline{\mu} - 2\bar{\beta}_1M_3 > 0, \\ \underline{\mu} - 2\bar{\beta}M_1 - 2\bar{\beta}_1M_2 > 0, \end{cases} \quad (4.1)$$

where  $M_1, M_2, M_3$  is given in (3.1), then let  $X(t) = (S_1(t), V_1(t), I_1(t), R_1(t))$  and  $Y(t) = (S_2(t), V_2(t), I_2(t), R_2(t))$  are any two positive solutions of the system (1.2), we have

$$\lim_{t \rightarrow \infty} |X(t) - Y(t)| = 0.$$

*Proof.* From Theorem 1, it follows that for  $\epsilon = \max \epsilon_i (i = 1, 2, \dots, 8)$  and  $T \geq T_8$  such that

$$\begin{aligned} m_1 - \epsilon < S_j < M_1 + \epsilon, \quad m_2 - \epsilon < V_j < M_2 + \epsilon, \\ m_3 - \epsilon < I_j < M_3 + \epsilon, \quad m_4 - \epsilon < R_j < M_4 + \epsilon, \end{aligned} \quad (4.2)$$

for all  $t \geq T$  and  $j = 1, 2$ , consider the following Lyapunov function

$$\tilde{V}(t) = \tilde{V}_1(t) + \tilde{V}_2(t) + \tilde{V}_3(t) + \tilde{V}_4(t), \quad (4.3)$$

where

$$\begin{aligned} \tilde{V}_1(t) &= |S_1(t) - S_2(t)|, & \tilde{V}_2(t) &= |V_1(t) - V_2(t)|, \\ \tilde{V}_3(t) &= |I_1(t) - I_2(t)|, & \tilde{V}_4(t) &= |R_1(t) - R_2(t)|. \end{aligned}$$

We define a function  $\sigma(\varphi(t))$  in the following way, the function  $\varphi(t)$  is supposed to be a real valued scalar function,

$$\sigma(\varphi(t)) = \begin{cases} 1, & \text{if } \varphi(t) > 0; \text{ or if } \varphi(t) = 0 \text{ and } \varphi'(t) > 0, \\ 0, & \text{if } \varphi(t) = 0 \text{ and } \varphi'(t) = 0, \\ -1, & \text{if } \varphi(t) < 0; \text{ or if } \varphi(t) = 0 \text{ and } \varphi'(t) < 0. \end{cases}$$

Then, it can be obtained that  $|\varphi(t)| = \varphi(t)\sigma(\varphi(t))$  and  $D^+|\varphi(t)| = \varphi'(t)\sigma(\varphi(t))$  where  $D^+$  denotes a right hand Dini derivative, it follows that

$$\begin{aligned} & D^+ \tilde{V}_1(t) \\ &= \sigma(S_1 - S_2)(S'_1 - S'_2) \\ &= \sigma(S_1 - S_2)[- \beta(t)S_1I_1 - (\mu(t) + \alpha(t))S_1 + \beta(t)S_2I_2 + (\mu(t) + \alpha(t))S_2] \\ &= -(\mu(t) + \alpha(t))|S_1 - S_2| - \sigma(S_1 - S_2)\beta(t)(S_1I_1 - S_1I_2 + S_1I_2 - S_2I_2) \\ &= -(\mu(t) + \alpha(t))|S_1 - S_2| - \sigma(S_1 - S_2)\beta(t)I_2(S_1 - S_2) \\ &\quad - \sigma(S_1 - S_2)\beta(t)S_1(I_1 - I_2) \\ &\leq -(\mu(t) + \alpha(t))|S_1 - S_2| + \beta(t)I_2|S_1 - S_2| + \beta(t)S_1|I_1 - I_2|. \end{aligned}$$



Similarly,

$$\begin{aligned}
 & D^+ \tilde{V}_2(t) \\
 &= \sigma(V_1 - V_2)(V'_1 - V'_2) \\
 &= \sigma(V_1 - V_2)[\alpha(t)S_1 - \beta_1(t)V_1I_1 - (\mu(t) + \gamma_1(t))(V_1 - V_2) - \alpha(t)S_2 + \beta_1(t)V_2I_2] \\
 &= -(\mu(t) + \gamma_1(t))|V_1 - V_2| + \sigma(V_1 - V_2)\alpha(t)(S_1 - S_2) \\
 &\quad - \sigma(V_1 - V_2)\beta_1(t)(V_1I_1 - V_2I_1 + V_2I_1 - V_2I_2) \\
 &= -(\mu(t) + \gamma_1(t))|V_1 - V_2| + \sigma(V_1 - V_2)\alpha(t)(S_1 - S_2) - \sigma(V_1 - V_2)\beta_1(t)I_1(V_1 - V_2) \\
 &\quad - \sigma(V_1 - V_2)\beta_1(t)V_2(I_1 - I_2) \\
 &\leq -(\mu(t) + \gamma_1(t))|V_1 - V_2| + \alpha(t)|S_1 - S_2| + \beta_1(t)I_1|V_1 - V_2| + \beta_1(t)V_2|I_1 - I_2|,
 \end{aligned}$$

$$\begin{aligned}
 & D^+ \tilde{V}_3(t) \\
 &= \sigma(I_1 - I_2)(I'_1 - I'_2) \\
 &= \sigma(I_1 - I_2)[\beta(t)S_1I_1 + \beta_1(t)V_1I_1 - (\gamma(t) + \mu(t))I_1(t) - \beta(t)S_2I_2 \\
 &\quad - \beta_1(t)V_2I_2 + (\gamma(t) + \mu(t))I_2(t) - \beta(t)S_1I_2 + \beta(t)S_1I_2 - \beta_1(t)V_2I_1 \\
 &\quad + \beta_1(t)V_2I_1 + k(t)(R_1 - R_2)] \\
 &= -\sigma(I_1 - I_2)(\gamma(t) + \mu(t))(I_1 - I_2) + \sigma(I_1 - I_2)[\beta_1(t)V_1I_1 - \beta_1(t)V_2I_2 \\
 &\quad - \beta_1(t)V_2I_1 + \beta_1(t)V_2I_1] + \sigma(I_1 - I_2)[\beta(t)S_1I_1 - \beta(t)S_2I_2 - \beta(t)S_1I_2 \\
 &\quad + \beta(t)S_1I_2] + \sigma(I_1 - I_2)k(t)(R_1 - R_2) \\
 &= -(\gamma(t) + \mu(t))|I_1 - I_2| + \sigma(I_1 - I_2)\beta_1(t)I_1(V_1 - V_2) + \sigma(I_1 - I_2)\beta_1(t) \\
 &\quad \times V_2(I_1 - I_2) + \sigma(I_1 - I_2)\beta(t)S_1(I_1 - I_2) + \sigma(I_1 - I_2)\beta(t)I_2(S_1 - S_2) \\
 &\quad + \sigma(I_1 - I_2)k(t)(R_1 - R_2) \\
 &\leq -(\gamma(t) + \mu(t))|I_1 - I_2| + \beta_1(t)I_1|V_1 - V_2| + \beta_1(t)V_2|I_1 - I_2| \\
 &\quad + \beta(t)S_1|I_1 - I_2| + \beta(t)I_2|S_1 - S_2| + k(t)|R_1 - R_2|,
 \end{aligned}$$

$$\begin{aligned}
 & D^+ \tilde{V}_4(t) \\
 &= \sigma(R_1 - R_2)(R'_1 - R'_2) \\
 &= \sigma(R_1 - R_2)[\gamma_1(t)V_1 + \gamma(t)I_1 - (\mu(t) + k(t))R_1 - \gamma_1(t)V_2 - \gamma(t)I_2 \\
 &\quad + (\mu(t) + k(t))R_2] \\
 &= \sigma(R_1 - R_2)\gamma_1(t)(V_1 - V_2) + \sigma(R_1 - R_2)\gamma(t)(I_1 - I_2) - \sigma(R_1 - R_2) \\
 &\quad \times (\mu(t) + k(t))(R_1 - R_2) \\
 &\leq \gamma_1(t)|V_1 - V_2| + \gamma(t)|I_1 - I_2| - (\mu(t) + k(t))|R_1 - R_2|.
 \end{aligned}$$

For  $t > T$ , we have

$$\begin{aligned}
 & D^+ \tilde{V}(t) \\
 &\leq -(\mu(t) - 2\beta(t)I_2)|S_1 - S_2| - (\mu(t) - 2\beta_1(t)I_1)|V_1 - V_2| \\
 &\quad - (\mu(t) - 2\beta(t)S_1 - 2\beta_1(t)V_2)|I_1 - I_2| - \mu(t)|R_1 - R_2| \\
 &\leq -[\underline{\mu} - 2\bar{\beta}(M_3 + \epsilon)]|S_1 - S_2| - [\underline{\mu} - 2\bar{\beta}_1(M_3 + \epsilon)]|V_1 - V_2| \\
 &\quad - [\underline{\mu} - 2\bar{\beta}(M_1 + \epsilon) - 2\bar{\beta}_1(M_2 + \epsilon)]|I_1 - I_2| - \underline{\mu}|R_1 - R_2|.
 \end{aligned}$$

When we choose  $\epsilon \rightarrow 0$ , the above relation still holds. Define  $\phi = \min\{\underline{\mu} - 2\bar{\beta}M_3, \underline{\mu}, \underline{\mu} - 2\bar{\beta}_1M_3, \underline{\mu} - 2\bar{\beta}M_1 - 2\bar{\beta}_1M_2\}$ . The above inequation takes the following form

$$D^+ \tilde{V}(t) \leq -\phi [ |S_1 - S_2| + |V_1 - V_2| + |I_1 - I_2| + |R_1 - R_2| ].$$

Integrating the above inequation from  $T$  to  $t$ , we have

$$\tilde{V}(t) + \phi \int_T^t \tilde{V}(y) dy \leq \tilde{V}(T) < +\infty. \quad (4.4)$$

It can be obtained from (1.2), (4.2), (4.3) that  $\tilde{V}(t)$  is uniformly continuous on  $(T, +\infty)$ . Then it can be obtained by Lemma 2 that  $\tilde{V}(t) \rightarrow 0$  as  $t \rightarrow +\infty$ . Otherwise,  $\phi \int_T^t \tilde{V}(y) dy \rightarrow +\infty$  when  $t \rightarrow \infty$  which is in contradiction with (4.4), then we get that

$$\begin{aligned} \lim_{t \rightarrow \infty} |S_1(t) - S_2(t)| &= 0, & \lim_{t \rightarrow \infty} |V_1(t) - V_2(t)| &= 0, \\ \lim_{t \rightarrow \infty} |I_1(t) - I_2(t)| &= 0, & \lim_{t \rightarrow \infty} |R_1(t) - R_2(t)| &= 0. \end{aligned}$$

□

**Theorem 3.** *Suppose all the conditions of Theorem 2 hold, then system (1.2) admits a unique almost periodic solution, which is global attractive. As a result, any solution of (1.2) is asymptotically almost periodic.*

*Proof.* For convenience, let  $\mathfrak{F} = \{\mu(t), \alpha(t), \beta(t), \beta_1(t), \gamma(t), \gamma_1(t), k(t)\}$  and

$$T(\mathfrak{F}, \epsilon) = \bigcap_{f \in \mathfrak{F}} T(f, \epsilon),$$

where  $T(f, \epsilon)$  is the set of  $\epsilon$ -almost periods for  $f$ . Since  $\mu(t), \alpha(t), \beta(t), \beta_1(t), \gamma(t), \gamma_1(t), k(t)$  are almost periodic functions, there exists a sequence  $\{t_n\}$  with  $t_n \rightarrow \infty$  as  $n \rightarrow \infty$ , such that

$$\lim_{n \rightarrow \infty} f(t + t_n) = f(t), \quad f \in \mathfrak{F}, \quad t \in \mathbb{R}. \quad (4.5)$$

Let

$$Q(t) = (q_1(t), q_2(t), q_3(t), q_4(t)), \quad t \geq 0,$$

be a bounded positive solution of the model (1.2). Then from (4.5), we can obtain that  $Q(t + t_n)$  is the solution of the following system for  $t \in \mathbb{R}^+$ :

$$\begin{cases} \frac{dS(t)}{dt} = \mu(t + t_n) - \beta(t + t_n)S(t)I(t) - (\mu(t + t_n) + \alpha(t + t_n))S(t), \\ \frac{dV(t)}{dt} = \alpha(t + t_n)S(t) - \beta_1(t + t_n)V(t)I(t) - (\mu(t + t_n) + \gamma_1(t + t_n))V(t), \\ \frac{dI(t)}{dt} = \beta(t + t_n)S(t)I(t) + \beta_1(t + t_n)V(t)I(t) - \gamma(t + t_n)I(t) \\ \quad - \mu(t + t_n)I(t) + k(t + t_n)R(t), \\ \frac{dR(t)}{dt} = \gamma_1(t + t_n)V(t) + \gamma(t + t_n)I(t) - \mu(t + t_n)R(t) - k(t + t_n)R(t). \end{cases} \quad (4.6)$$

Let  $\tau^* \in \mathbb{R}^+ \cap T(\mathfrak{F}, \varepsilon)$ , thus

$$|f(t + \tau^*) - f(t)| < \varepsilon, f \in \mathfrak{F}, t \in \mathbb{R}. \quad (4.7)$$

Define  $W(t)$  as follows

$$W(t) = W_1(t) + W_2(t) + W_3(t) + W_4(t), t \geq 0, \quad (4.8)$$

where

$$W_i(t) = |q_i(t + \tau^*) - q_i(t)| \quad (i = 1, 2, 3, 4).$$

Since the  $Q(t)$  is the solution of system (1.2), it can be obtained that

$$\begin{aligned} & D^+ W_1(t) \\ &= \sigma(q_1(t + \tau^*) - q_1(t))(q_1'(t + \tau^*) - q_1'(t)) \\ &= \sigma(q_1(t + \tau^*) - q_1(t))[\mu(t + \tau^*) - \beta(t + \tau^*)q_1(t + \tau^*)q_3(t + \tau^*) - \mu(t) \\ &\quad - (\mu(t + \tau^*) + \alpha(t + \tau^*))q_1(t + \tau^*) + \beta(t)q_1(t)q_3(t) + (\mu(t) + \alpha(t))q_1(t)] \\ &= \sigma(q_1(t + \tau^*) - q_1(t))(\mu(t + \tau^*) - \mu(t)) + \sigma(q_1(t + \tau^*) - q_1(t)) \\ &\quad \times [-\beta(t + \tau^*)q_1(t + \tau^*)q_3(t + \tau^*) + \beta(t + \tau^*)q_1(t)q_3(t) - \beta(t + \tau^*) \\ &\quad \times q_1(t)q_3(t) + \beta(t)q_1(t)q_3(t) - (\mu(t + \tau^*) + \alpha(t + \tau^*))q_1(t + \tau^*) \\ &\quad + (\mu(t + \tau^*) + \alpha(t + \tau^*))q_1(t) - (\mu(t + \tau^*) + \alpha(t + \tau^*))q_1(t) \\ &\quad + (\mu(t) + \alpha(t))q_1(t)] \\ &\leq |\mu(t + \tau^*) - \mu(t)| + \sigma(q_1(t + \tau^*) - q_1(t))[-\beta(t + \tau^*)q_1(t + \tau^*)q_3(t + \tau^*) \\ &\quad + \beta(t + \tau^*)q_1(t)q_3(t)] + q_1(t)q_3(t)|\beta(t) - \beta(t + \tau^*)| \\ &\quad - (\mu(t + \tau^*) + \alpha(t + \tau^*))|q_1(t + \tau^*) - q_1(t)| \\ &\quad + q_1(t)|\mu(t) - \mu(t + \tau^*) + \alpha(t) - \alpha(t + \tau^*)| \\ &\leq \varepsilon - \beta(t + \tau^*)q_3(t + \tau^*)|q_1(t + \tau^*) - q_1(t)| - \beta(t + \tau^*)q_1(t)|q_3(t + \tau^*) - q_3(t)| \\ &\quad - (\mu(t + \tau^*) + \alpha(t + \tau^*))|q_1(t + \tau^*) - q_1(t)| + 2\varepsilon q_1(t) + \varepsilon q_1(t)q_3(t), \end{aligned}$$

$$\begin{aligned} & D^+ W_2(t) \\ &= \sigma(q_2(t + \tau^*) - q_2(t))(q_2'(t + \tau^*) - q_2'(t)) \\ &= \sigma(q_2(t + \tau^*) - q_2(t))[\alpha(t + \tau^*)q_1(t + \tau^*) - \beta_1(t + \tau^*)q_2(t + \tau^*)q_3(t + \tau^*) \\ &\quad - (\mu(t + \tau^*) + \gamma_1(t + \tau^*))q_2(t + \tau^*) - \alpha(t)q_1(t) + \beta_1(t)q_2(t)q_3(t) \\ &\quad + (\mu(t) + \gamma_1(t))q_2(t)] \\ &= \sigma(q_2(t + \tau^*) - q_2(t))[\alpha(t + \tau^*)q_1(t + \tau^*) - \alpha(t + \tau^*)q_1(t) + \alpha(t + \tau^*)q_1(t) \\ &\quad - \alpha(t)q_1(t) - \beta_1(t + \tau^*)q_2(t + \tau^*)q_3(t + \tau^*) + \beta_1(t + \tau^*)q_2(t)q_3(t) \\ &\quad - \beta_1(t + \tau^*)q_2(t)q_3(t) + \beta_1(t)q_2(t)q_3(t) - (\mu(t + \tau^*) + \gamma_1(t + \tau^*)) \\ &\quad \times q_2(t + \tau^*) + (\mu(t + \tau^*) + \gamma_1(t + \tau^*))q_2(t) - (\mu(t + \tau^*) + \gamma_1(t + \tau^*))q_2(t) \\ &\quad + (\mu(t) + \gamma_1(t))q_2(t)] \\ &\leq \alpha(t + \tau^*)|q_1(t + \tau^*) - q_1(t)| + q_1(t)|\alpha(t + \tau^*) - \alpha(t)| + q_2(t)q_3(t) \\ &\quad \times |\beta_1(t) - \beta_1(t + \tau^*)| - \sigma(q_2(t + \tau^*) - q_2(t))\beta_1(t + \tau^*)[q_2(t + \tau^*)q_3(t + \tau^*) \\ &\quad - q_2(t)q_3(t + \tau^*) + q_2(t)q_3(t + \tau^*) - q_2(t)q_3(t)] - (\mu(t + \tau^*) + \gamma_1(t + \tau^*)) \end{aligned}$$

$$\begin{aligned}
& \times |q_2(t + \tau^*) - q_2(t)| + q_2(t)|\mu(t) - \mu(t + \tau^*) + \gamma_1(t) - \gamma_1(t + \tau^*)| \\
\leq & \alpha(t + \tau^*)|q_1(t + \tau^*) - q_1(t)| - (\beta_1(t + \tau^*)q_3(t + \tau^*) + \mu(t + \tau^*) \\
& + \gamma_1(t + \tau^*))|q_2(t + \tau^*) - q_2(t)| - \beta_1(t + \tau^*)q_2(t)|q_3(t + \tau^*) - q_3(t)| \\
& + q_1(t)\varepsilon + q_2(t)q_3(t)\varepsilon + 2q_2(t)\varepsilon,
\end{aligned}$$

$$\begin{aligned}
& D^+ W_3(t) \\
= & \sigma(q_3(t + \tau^*) - q_3(t))(q_3'(t + \tau^*) - q_3'(t)) \\
= & \sigma(q_3(t + \tau^*) - q_3(t))[\beta(t + \tau^*)q_1(t + \tau^*)q_3(t + \tau^*) + \beta_1(t + \tau^*) \\
& \times q_2(t + \tau^*)q_3(t + \tau^*) - (\gamma(t + \tau^*) + \mu(t + \tau^*))q_3(t + \tau^*) + k(t + \tau^*) \\
& \times q_4(t + \tau^*) - \beta(t)q_1(t)q_3(t) - \beta_1(t)q_2(t)q_3(t) + (\gamma(t) + \mu(t))q_3(t) - k(t)q_4(t)] \\
= & \sigma(q_3(t + \tau^*) - q_3(t))[\beta(t + \tau^*)q_1(t + \tau^*)q_3(t + \tau^*) - \beta(t + \tau^*)q_1(t)q_3(t) \\
& + \beta(t + \tau^*)q_1(t)q_3(t) - \beta(t)q_1(t)q_3(t)] + \sigma(q_3(t + \tau^*) - q_3(t)) \\
& \times [\beta_1(t + \tau^*)q_2(t + \tau^*)q_3(t + \tau^*) - \beta_1(t + \tau^*)q_2(t)q_3(t) + \beta_1(t + \tau^*)q_2(t)q_3(t) \\
& - \beta_1(t)q_2(t)q_3(t)] + \sigma(q_3(t + \tau^*) - q_3(t))[-(\gamma(t + \tau^*) + \mu(t + \tau^*))q_3(t + \tau^*) \\
& + (\gamma(t + \tau^*) + \mu(t + \tau^*))q_3(t) - (\gamma(t + \tau^*) + \mu(t + \tau^*))q_3(t) \\
& + (\gamma(t) + \mu(t))q_3(t)] + \sigma(q_3(t + \tau^*) - q_3(t))[k(t + \tau^*)q_4(t + \tau^*) \\
& - k(t + \tau^*)q_4(t) + k(t + \tau^*)q_4(t) - k(t)q_4(t)] \\
\leq & q_1(t)q_3(t)|\beta(t + \tau^*) - \beta(t)| + \sigma(q_3(t + \tau^*) - q_3(t))\beta(t + \tau^*)[q_1(t + \tau^*) \\
& \times q_3(t + \tau^*) - q_1(t)q_3(t + \tau^*) + q_1(t)q_3(t + \tau^*) - q_1(t)q_3(t)] \\
& + q_2(t)q_3(t)|\beta_1(t + \tau^*) - \beta_1(t)| + \sigma(q_3(t + \tau^*) - q_3(t)) \\
& \times \beta_1(t + \tau^*)[q_2(t + \tau^*)q_3(t + \tau^*) - q_2(t)q_3(t + \tau^*) \\
& + q_2(t)q_3(t + \tau^*) - q_2(t)q_3(t)] - (\gamma(t + \tau^*) + \mu(t + \tau^*))|q_3(t + \tau^*) - q_3(t)| \\
& + q_3(t)|\gamma(t) + \mu(t) - \gamma(t + \tau^*) - \mu(t + \tau^*)| + k(t + \tau^*)|q_4(t + \tau^*) - q_4(t)| \\
& + q_4(t)|k(t + \tau^*) - k(t)| \\
\leq & \beta(t + \tau^*)q_3(t + \tau^*)|q_1(t + \tau^*) - q_1(t)| + \beta_1(t + \tau^*)q_3(t + \tau^*) \\
& \times |q_2(t + \tau^*) - q_2(t)| + (\beta(t + \tau^*)q_1(t) + \beta_1(t + \tau^*)q_2(t) - \gamma(t + \tau^*) \\
& - \mu(t + \tau^*))|q_3(t + \tau^*) - q_3(t)| + k(t + \tau^*)|q_4(t + \tau^*) - q_4(t)| + \varepsilon q_1(t)q_3(t) \\
& + \varepsilon q_2(t)q_3(t) + 2\varepsilon q_3(t) + \varepsilon q_4(t),
\end{aligned}$$

$$\begin{aligned}
& D^+ W_4(t) \\
= & \sigma(q_4(t + \tau^*) - q_4(t))(q_4'(t + \tau^*) - q_4'(t)) \\
= & \sigma(q_4(t + \tau^*) - q_4(t))[\gamma_1(t + \tau^*)q_2(t + \tau^*) + \gamma(t + \tau^*)q_3(t + \tau^*) \\
& - \mu(t + \tau^*)q_4(t + \tau^*) - k(t + \tau^*)q_4(t + \tau^*) - \gamma_1(t)q_2(t) - \gamma(t)q_3(t) \\
& + \mu(t)q_4(t) + k(t)q_4(t)] \\
= & \sigma(q_4(t + \tau^*) - q_4(t))[\gamma_1(t + \tau^*)q_2(t + \tau^*) - \gamma_1(t + \tau^*)q_2(t) \\
& + \gamma_1(t + \tau^*)q_2(t) - \gamma_1(t)q_2(t)] + \sigma(q_4(t + \tau^*) - q_4(t))[\gamma(t + \tau^*)q_3(t + \tau^*)
\end{aligned}$$

$$\begin{aligned}
& -\gamma(t+\tau^*)q_3(t) + \gamma(t+\tau^*)q_3(t) - \gamma(t)q_3(t) + \sigma(q_4(t+\tau^*) - q_4(t)) \\
& \times [-\mu(t+\tau^*)q_4(t+\tau^*) + \mu(t+\tau^*)q_4(t) - \mu(t+\tau^*)q_4(t) + \mu(t)q_4(t)] \\
& + \sigma(q_4(t+\tau^*) - q_4(t))[-k(t+\tau^*)q_4(t+\tau^*) + k(t+\tau^*)q_4(t) \\
& - k(t+\tau^*)q_4(t) + k(t)q_4(t)] \\
& \leq \gamma_1(t+\tau^*)|q_2(t+\tau^*) - q_2(t)| + q_2(t)|\gamma_1(t+\tau^*) - \gamma_1(t)| + \gamma(t+\tau^*) \\
& \quad \times |q_3(t+\tau^*) - q_3(t)| + q_3(t)|\gamma(t+\tau^*) - \gamma(t)| - (\mu(t+\tau^*) + k(t+\tau^*)) \\
& \quad \times |q_4(t+\tau^*) - q_4(t)| + q_4(t)|\mu(t) - \mu(t+\tau^*)| + q_4(t)|k(t) - k(t+\tau^*)| \\
& \leq \gamma_1(t+\tau^*)|q_2(t+\tau^*) - q_2(t)| + \gamma(t+\tau^*)|q_3(t+\tau^*) - q_3(t)| \\
& \quad - (\mu(t+\tau^*) + k(t+\tau^*))|q_4(t+\tau^*) - q_4(t)| + (q_2(t) + q_3(t) + 2q_4(t))\varepsilon.
\end{aligned}$$

Then, let  $M = \max\{M_i\}(i = 1, 2, 3, 4)$ , we obtain

$$\begin{aligned}
& D^+W(t) \\
& = -\mu(t+\tau^*)|q_1(t+\tau^*) - q_1(t)| - \mu(t+\tau^*)|q_2(t+\tau^*) - q_2(t)| \\
& \quad - \mu(t+\tau^*)|q_3(t+\tau^*) - q_3(t)| - \mu(t+\tau^*)|q_4(t+\tau^*) - q_4(t)| + \varepsilon \\
& \quad + 2\varepsilon q_1(t)q_3(t) + 3\varepsilon q_1(t) + 2\varepsilon q_2(t)q_3(t) + 3\varepsilon q_2(t) + 3\varepsilon q_3(t) + 3\varepsilon q_4(t) \\
& \leq -\underline{\mu}W(t) + 12\varepsilon M + 4\varepsilon M^2 + \varepsilon.
\end{aligned} \tag{4.9}$$

Integrating both sides of (4.9) from  $t$  to  $T$  ( $T > T_8$ ), and let  $\tilde{\varepsilon} = \frac{24\varepsilon M + 8\varepsilon M^2 + 2\varepsilon}{\underline{\mu}}$  we get

$$\begin{aligned}
W(t) & \leq \frac{12\varepsilon M + 4\varepsilon M^2 + \varepsilon}{\underline{\mu}} + \left( W(T) - \frac{12\varepsilon M + 4\varepsilon M^2 + \varepsilon}{\underline{\mu}} \right) e^{-\underline{\mu}t} \\
& = \frac{1}{2}\tilde{\varepsilon} + \left( W(T) - \frac{12\varepsilon M + 4\varepsilon M^2 + \varepsilon}{\underline{\mu}} \right) e^{-\underline{\mu}t}.
\end{aligned}$$

It can be obtained that  $\lim_{t \rightarrow \infty} \left( W(T) - \frac{12\varepsilon M + 4\varepsilon M^2 + \varepsilon}{\underline{\mu}} \right) e^{-\underline{\mu}t} = 0$ . Then there exists  $\tilde{T} > T$  such that for all  $t > \tilde{T}$ ,

$$\left| \left( W(T) - \frac{12\varepsilon M + 4\varepsilon M^2 + \varepsilon}{\underline{\mu}} \right) e^{-\underline{\mu}t} \right| < \frac{1}{2}\tilde{\varepsilon}.$$

Thus, we get that

$$W(t) < \tilde{\varepsilon}.$$

Then, in view of (4.8), for all  $t > \tilde{T}$ ,

$$|q_i(t+\tau^*) - q_i(t)| < \tilde{\varepsilon}, \quad i = 1, 2, 3, 4.$$

Therefore,  $\tau^* \in T^+(q_i, \tilde{\varepsilon})$ , which means

$$T(\mathfrak{I}, \varepsilon) \cap \mathbb{R}^+ \subset T^+(q_i, \tilde{\varepsilon}).$$

Thus, according to the Proposition 1,  $T^+(q_i, \mathcal{E})$  is relatively dense in  $\mathbb{R}^+$  and  $q_i(t)$  is the asymptotically almost periodic solution of (1.2). Then there is an almost periodic function  $q_{i1}(t)$  defined on  $\mathbb{R}$  and a continuous function  $q_{i2}(t)$  defined on  $\mathbb{R}^+$  with  $\lim_{t \rightarrow \infty} q_{i2}(t) = 0$ , such that

$$q_i(t) = q_{i1}(t) + q_{i2}(t), \quad t \in \mathbb{R}^+.$$

We denote

$$Q(t) = Q^{ap}(t) + Q^e(t), \quad t \in \mathbb{R}^+,$$

where  $Q^{ap}(t) = (q_{11}(t), q_{21}(t), q_{31}(t), q_{41}(t))$ ,  $Q^e(t) = (q_{12}(t), q_{22}(t), q_{32}(t), q_{42}(t))$ .

Now we prove that  $Q^{ap}(t)$  is an almost periodic solution of the system (1.2) for  $t \in \mathbb{R}$ . Since  $q_{i1}(t)$  is almost periodic function, there exist a sequence  $\{t_n\}$  with  $t_n \rightarrow \infty$  as  $n \rightarrow \infty$  such that

$$\lim_{n \rightarrow \infty} q_{i1}(t + t_n) = q_{i1}(t), \quad t \in \mathbb{R},$$

and

$$\lim_{n \rightarrow \infty} q_{i2}(t + t_n) = 0, \quad t \in \mathbb{R}^+.$$

Then we get  $\lim_{n \rightarrow \infty} q_i(t + t_n) = q_i(t)$  for  $t \in \mathbb{R}^+$ ,  $i = 1, 2, 3, 4$ . Moreover,

$$\begin{aligned} & \lim_{n \rightarrow \infty} \dot{q}_i(t + t_n) \\ &= \lim_{n \rightarrow \infty} \lim_{h \rightarrow 0} \frac{q_i(t + t_n + h) - q_i(t + t_n)}{h} \\ &= \lim_{h \rightarrow 0} \lim_{n \rightarrow \infty} \frac{q_i(t + t_n + h) - q_i(t + t_n)}{h} \\ &= \lim_{h \rightarrow 0} \frac{q_{i1}(t + t_n + h) - q_{i1}(t + t_n)}{h} \\ &= \dot{q}_{i1}. \end{aligned}$$

From (4.5), for  $t \in \mathbb{R}$ , we get

$$\begin{aligned} \dot{q}_{11}(t) &= \lim_{n \rightarrow \infty} \dot{q}_1(t + t_n) \\ &= \lim_{n \rightarrow \infty} [\mu(t + t_n) - \beta(t + t_n)q_1(t + t_n)q_3(t + t_n) \\ &\quad - (\mu(t + t_n) + \alpha(t + t_n))q_1(t + t_n)] \\ &= \mu(t) - \beta(t)q_{11}(t)q_{31}(t) - (\mu(t) + \alpha(t))q_{11}(t). \end{aligned}$$

Similarly, for  $t \in \mathbb{R}$ ,

$$\begin{aligned} \dot{q}_{21}(t) &= \lim_{n \rightarrow \infty} \dot{q}_2(t + t_n) \\ &= \alpha(t)q_{11}(t) - \beta_1(t)q_{21}(t)q_{31}(t) - (\mu(t) + \gamma_1(t))q_{21}(t), \\ \dot{q}_{31}(t) &= \lim_{n \rightarrow \infty} \dot{q}_3(t + t_n) \\ &= \beta(t)q_{11}(t)q_{31}(t) + \beta_1(t)q_{21}(t)q_{31}(t) - \gamma(t)q_{31}(t) - \mu(t)q_{31}(t) + k(t)q_{41}(t), \\ \dot{q}_{41}(t) &= \lim_{n \rightarrow \infty} \dot{q}_4(t + t_n) \\ &= \gamma_1(t)q_{21}(t) + \gamma(t)q_{31}(t) - (\mu(t) + k(t))q_{41}(t). \end{aligned}$$

Thus, we get that  $Q^{ap}(t)$  is an almost periodic solution of system (1.2) for  $t \in \mathbb{R}$ .

Let  $H(t) = (S^*(t), V^*(t), I^*(t), R^*(t))$  be another solution of system (1.2). By Theorem 2, we get

$$\lim_{t \rightarrow \infty} |H(t) - Q^{ap}(t)| = 0,$$

which implies that the unique almost periodic solution  $Q^{ap}(t)$  is global attractive.  $\square$

## 5. Analysis of the recurrence rate $k$

In this section, we discuss the necessity of adding the recurrence rate  $k$  in this model. It is well known that the basic reproduction number  $R_0$  is the threshold value of the model, which shows that the disease persists or extinct in the population. Therefore, the relationship between  $R_0$  and  $k$  can be used to reflect the influence of  $k$  on this system. We consider the following autonomous model corresponding to model (1.2):

$$\begin{cases} \frac{dS(t)}{dt} = \mu - \beta S(t)I(t) - (\mu + \alpha)S(t), \\ \frac{dV(t)}{dt} = \alpha S(t) - \beta_1 V(t)I(t) - (\mu + \gamma_1)V(t), \\ \frac{dI(t)}{dt} = \beta S(t)I(t) + \beta_1 V(t)I(t) - \gamma I(t) - \mu I(t) + kR(t), \\ \frac{dR(t)}{dt} = \gamma_1 V(t) + \gamma I(t) - \mu R(t) - kR(t), \end{cases} \quad (5.1)$$

for  $t \in \mathbb{R}^+$ , where the parameters are all positive constants. It's easy to get the disease free equilibrium of the system (5.1),

$$E_0 = (S_0, V_0, I_0, R_0) = \left( \frac{\mu}{\mu + \alpha}, \frac{\alpha\mu}{(\mu + \gamma_1)(\mu + \alpha)}, 0, \frac{\gamma_1\alpha\mu}{(\mu + k)(\mu + \gamma_1)(\mu + \alpha)} \right).$$

Then let  $x = (I, V, R, S)^T$ , thus the system can be written as:

$$\frac{dx}{dt} = \mathcal{F}(x) - \mathcal{V}(x),$$

where

$$\mathcal{F}(x) = \begin{pmatrix} \beta S I + \beta_1 V I \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathcal{V}(x) = \begin{pmatrix} (\gamma + \mu)I - kR \\ \beta_1 V I + (\mu + \gamma_1)V - \alpha S \\ (\mu + k)R - \gamma_1 V - \gamma I \\ \beta S I + (\mu + \alpha)S - \mu \end{pmatrix}.$$

The Jacobian matrices of  $\mathcal{F}(x)$  and  $\mathcal{V}(x)$  at the disease free equilibrium  $E_0$  are respectively,

$$D\mathcal{F}(E_0) = \begin{pmatrix} F_{3 \times 3} & A \\ 0 & 0 \end{pmatrix}, \quad D\mathcal{V}(E_0) = \begin{pmatrix} V_{3 \times 3} & B \\ C & \mu + \alpha \end{pmatrix},$$

where,

$$A = \begin{pmatrix} \beta I \\ 0 \\ 0 \end{pmatrix}, \quad B = \begin{pmatrix} 0 \\ -\alpha \\ 0 \end{pmatrix}, \quad C = (\beta S_0 \quad 0 \quad 0),$$

the non-negative matrix  $F$  for the appearance of the new disease and the matrix  $V$  for the transition terms are given by

$$F = \begin{pmatrix} \beta S_0 + \beta_1 V_0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, V = \begin{pmatrix} \gamma + \mu & 0 & -k \\ \beta_1 V_0 & \mu + \gamma_1 & 0 \\ -\gamma & -\gamma_1 & \mu + k \end{pmatrix}. \quad (5.2)$$

By the next-generation operator method [43], the basic reproduction number of model (5.1) is calculated

$$R_0 = \rho(FV^{-1}) \\ = \left( \frac{\beta\mu}{\mu + \alpha} + \frac{\beta_1\alpha\mu}{(\mu + \gamma_1)(\mu + \alpha)} \right) \frac{(\mu + \gamma_1)(\mu + k)}{(\mu + k)(\mu + \gamma)(\mu + \gamma_1) + k\gamma_1\beta_1 V_0 - k\gamma(\mu + \gamma_1)}.$$

To get the sensitivity of  $R_0$  to  $k$ , following Chitnis and Hyman [44], the normalised forward sensitivity index with respect to  $k$  is given by

$$A_k = \frac{\frac{\partial R_0}{R_0}}{\frac{\partial k}{k}} = \frac{k}{R_0} \frac{\partial R_0}{\partial k}. \quad (5.3)$$

**Remark 1.** *The sensitivity index is to assess the relative change in state variables when a parameter of the model changes. If  $A_k$  is positive, the value of the reproduction number will increase as  $k$  increases. Similarly, if  $A_k$  is negative, the value of the reproduction number will decrease as  $k$  increases. In next section, we will give some parameter values to calculate  $A_k$ , and verify our analysis by numerical simulations, see Figures 4 and 5.*

*For non-autonomous model (1.2), due to the complexity of the model, we use numerical simulation to reflect the impact of the recurrence rate on the model. See Figures 6 and 7 in the next chapter.*

## 6. Numerical simulation

In this section, to illustrate the analytic results obtained above, we have presented some simulations of system (1.2) and system (5.1).

### 6.1. Almost periodic solutions of system (1.2)

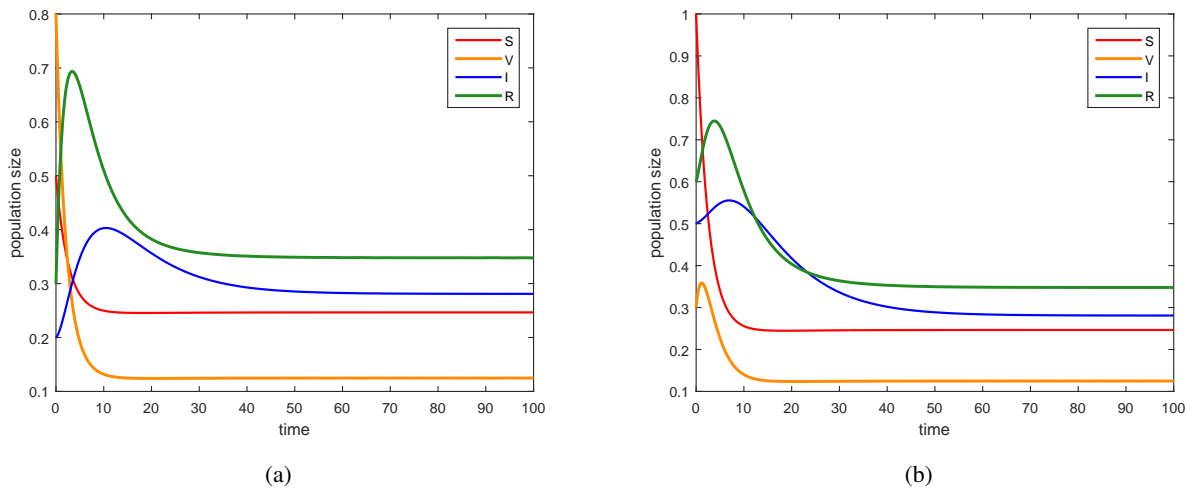
**Example 1.** *We choose parameter values as follows:*

$$\mu = 0.1, \beta = 0.02, \alpha = 0.3, \gamma_1 = 0.49, \gamma = 0.03, \beta_1 = 0.01, k = 0.1. \quad (6.1)$$

*Then we get that  $\underline{\mu} = \bar{\mu} = 0.1$ ,  $\underline{\beta} = \bar{\beta} = 0.02$ ,  $\underline{\alpha} = \bar{\alpha} = 0.3$ ,  $\underline{\gamma} = \bar{\gamma} = 0.03$ ,  $\underline{\gamma}_1 = \bar{\gamma}_1 = 0.49$ ,  $\underline{\beta}_1 = \bar{\beta}_1 = 0.01$ ,  $\underline{k} = \bar{k} = 0.1$ . All the sufficient conditions given in Theorem 2 for system (1.2) are well satisfied as*

$$\begin{cases} \underline{\mu} - 2\bar{\beta}M_3 = 0.06 > 0, \\ \underline{\mu} - 2\bar{\beta}_1M_3 = 0.08 > 0, \\ \underline{\mu} - 2\bar{\beta}M_1 - 2\bar{\beta}_1M_2 \approx 0.09 > 0. \end{cases}$$





**Figure 1.** Solution curves for the model (1.2) with parametric values in (6.1). (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

Using the parameter values in (6.1) to draw figures of the solution of model (1.2), we get Figure 1(a) and 1(b), and their corresponding initial values are  $(0.5, 0.8, 0.2, 0.3)$  and  $(1, 0.3, 0.5, 0.6)$ , respectively. It can be seen from 1 that as time goes by, all solutions tend to be a constant, which reflects the the model has a unique globally attractive positive almost periodic solution.

**Example 2.** We choose parameter values as follows:

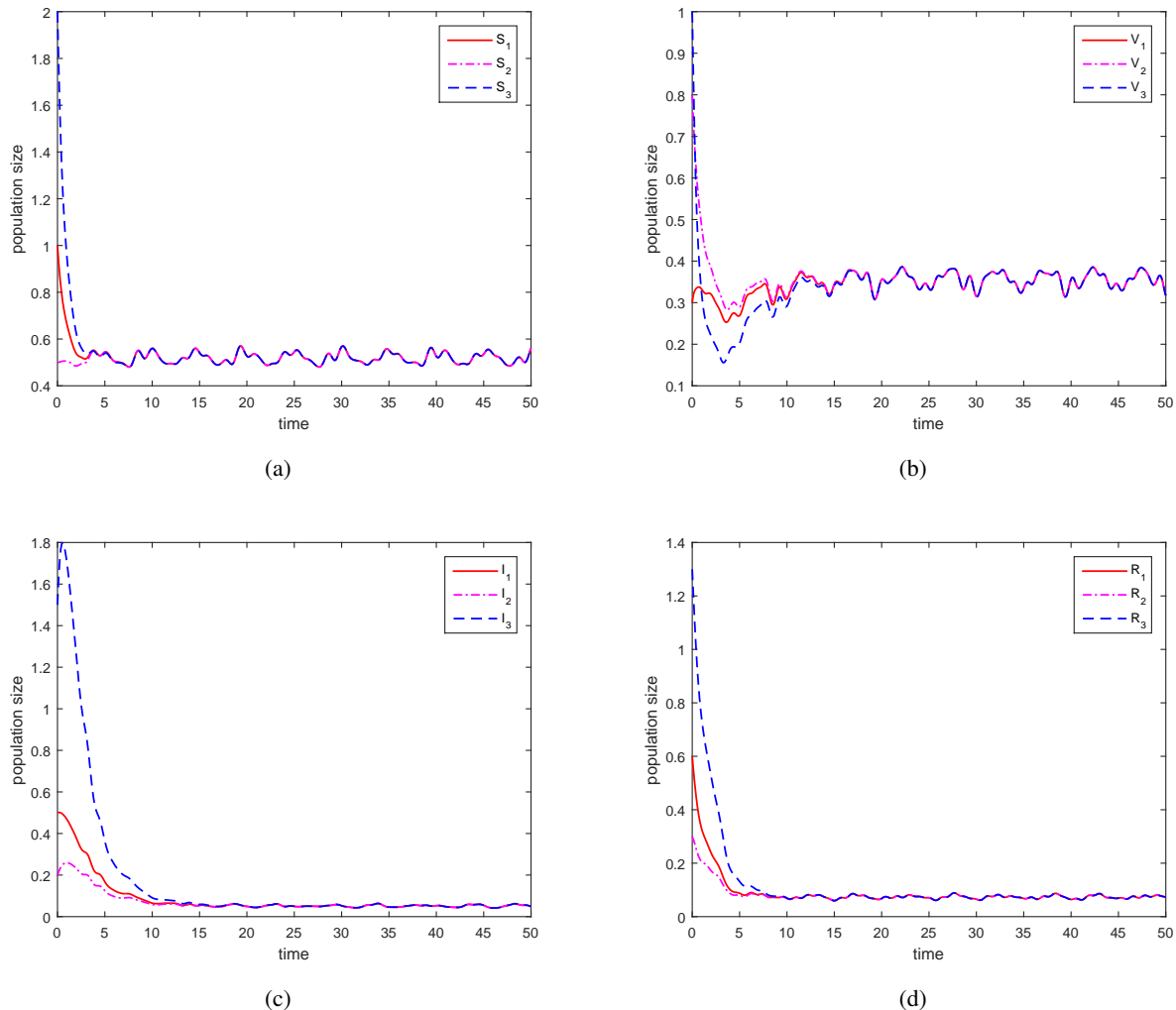
$$\begin{aligned}
 \mu &= 0.5 + \frac{1}{10}(\sin \sqrt{2}t + \cos \sqrt{7}t)^2, \beta = 0.02 + \frac{1}{40}(\sin \sqrt{2}t + \cos \sqrt{7}t)^2, \\
 \alpha &= 0.5 + \frac{1}{10}(\cos \sqrt{3}t)^2, \beta_1 = 0.01 + \frac{9}{400}(\sin \sqrt{3}t + \cos 3t)^2, \\
 \gamma_1 &= 0.1 + \frac{1}{10}(\cos \sqrt{7}t)^2, \gamma = 0.1 + \frac{1}{100}\cos 3t, \\
 k &= 0.2 + \frac{1}{10}\cos \sqrt{3}t.
 \end{aligned} \tag{6.2}$$

Then we get that  $\underline{\mu} = 0.5, \bar{\mu} = 0.9, \underline{\beta} = 0.02, \bar{\beta} = 0.12, \underline{\alpha} = 0.5, \bar{\alpha} = 0.6, \underline{\gamma} = 0.09, \bar{\gamma} = 0.11, \underline{\gamma}_1 = 0.1, \bar{\gamma}_1 = 0.5, \underline{\beta}_1 = 0.01, \bar{\beta}_1 = 0.1, \underline{k} = 0.1, \bar{k} = 0.3$ . All the sufficient conditions given in Theorem 2 for system (1.2) are well satisfied as

$$\begin{cases}
 \underline{\mu} - 2\bar{\beta}M_3 \approx 0.068 > 0, \\
 \underline{\mu} - 2\bar{\beta}_1M_3 = 0.14 > 0, \\
 \underline{\mu} - 2\bar{\beta}M_1 - 2\bar{\beta}_1M_2 = 0.14 > 0.
 \end{cases}$$

Then the model has a unique globally attractive positive almost periodic solution. Next, we use numerical simulations to verify our conclusion. The parameter values are shown in (6.2). For the convenience of observation, we have drawn 4 figures to reflect the curves of each solution when the initial values are different. It can be seen from Figure 2 that the solution is globally attractive. It can also be seen that

for different initial values, as time goes by,  $S_1, S_2, S_3; V_1, V_2, V_3; I_1, I_2, I_3; R_1, R_2, R_3$  tends to the same curve, respectively. The numerical simulation in Figures 2 and 3 strongly support the consequence.



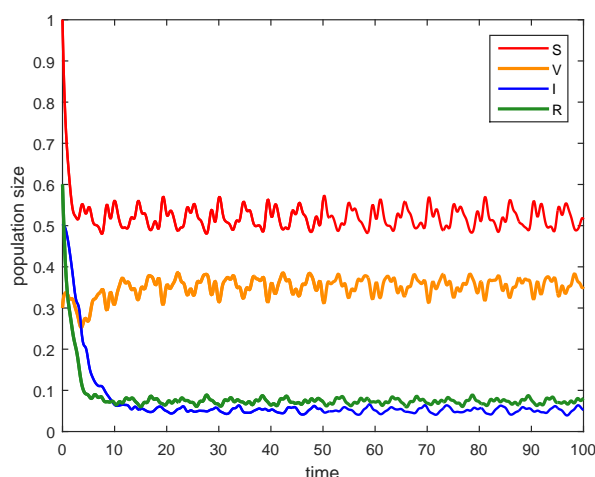
**Figure 2.** Numerical solutions of system (1.2) for the initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3), (1, 0.3, 0.5, 0.6), (2, 1, 1.5, 1.3)$ . (a) (b) (c) and (d) show that although the initial values are different, as time goes by,  $S_1, S_2, S_3; V_1, V_2, V_3; I_1, I_2, I_3; R_1, R_2, R_3$  tends to the same curve, respectively.

## 6.2. Simulation related to the recurrence rate $k$

For the autonomous system (5.1), from (5.3), take the parameter values as in (6.1), we get

$$A_k = \frac{k}{R_0} \frac{\partial R_0}{\partial k} \approx 0.06 > 0.$$

By Remark 1 in Section 5, we know that  $R_0$  will decrease as  $k$  decreases, which implies that reducing the recurrence rate  $k$  is very helpful to control the spread of infectious diseases.



**Figure 3.** Almost periodic solution of system (1.2) with parametric values in (6.2).

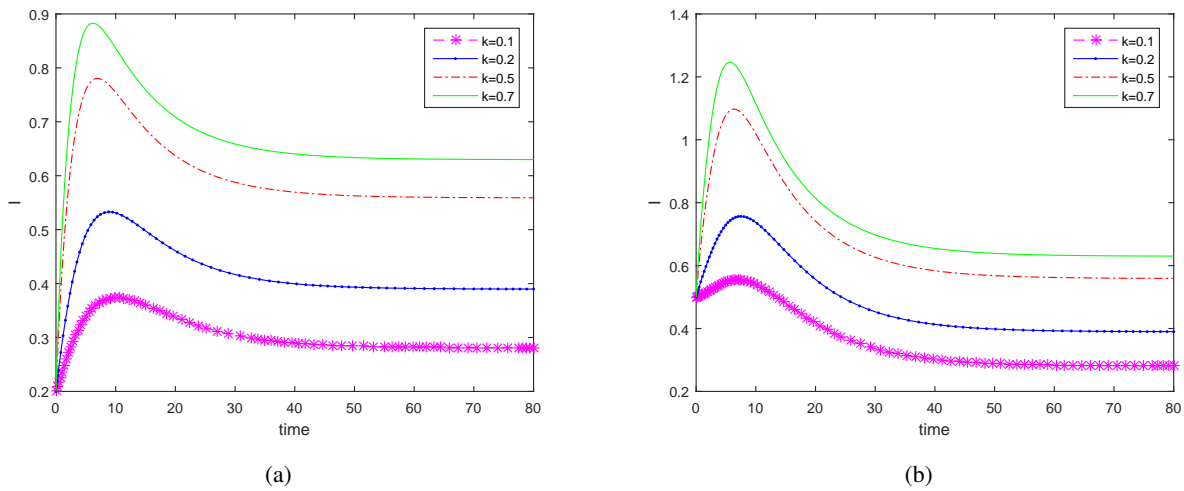
Next, we give numerical simulations to verify our views. Change the parameter  $k$  while keeping other parameters fixed as in (6.1). Figure 4 shows the effect of recurrence rate on the number of infectious people. It is easy to see that the values of  $k$  have a significant effect on the number of infectious people. As the value of  $k$  decrease, the value of  $I$  decrease. Figure 5 shows the effect of recurrence rate on the number of recovered people. It is easy to see that as the value of  $k$  decrease, the value of  $R$  increase. If we want to control an infectious disease, we certainly hope that the number of infectious people will decrease and the number of recovered people will increase. These two figures tell us that when  $k$  decreases, the disease can be well controlled, which is very consistent with our analysis.

For the nonautonomous system (1.2), we change the parameter  $k(t)$  while keeping other parameters fixed as in (6.2). It has similar conclusions with the autonomous model. Figures 6 and 7 show that when  $k$  decreases, the number of infectious people decreases, the number of recovered people increases, respectively. This also verifies that it is necessary to consider the recurrence rate in the model.

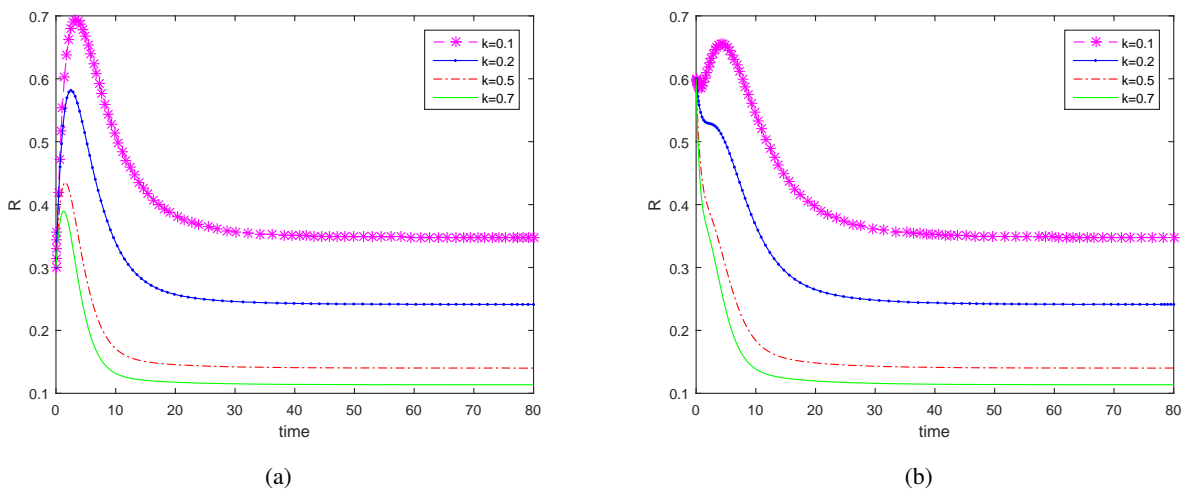
### 6.3. Simulation related to the vaccination rate $\alpha$

For the autonomous system (5.1), let  $\beta = 0.5$ , change the parameter  $\alpha$  while keeping other parameters fixed as in (6.1). Figure 8 shows the effect of vaccination rate on the number of infectious people. It is easy to see that the values of  $\alpha$  have a significant effect on the number of infectious people. As the value of  $\alpha$  increase, the value of  $I$  decrease. Figure 9 shows the effect of vaccination rate on the number of recovered people. It is easy to see that as the value of  $\alpha$  increase, the value of  $R$  increase. These two figures tell us that increasing the vaccination rate of the disease can control the spread of the disease.

For the nonautonomous system (1.2), let  $\beta(t) = 1.5 + \frac{1}{40}(\sin \sqrt{2}t + \cos \sqrt{7}t)^2$ , change the parameter  $\alpha(t)$  while keeping other parameters fixed as in (6.2). It has similar conclusions with the autonomous model. Figures 10 and 11 show that when  $\alpha(t)$  increases, the number of infectious people decreases, the number of recovered people increases, respectively. This shows that it is very meaningful to consider the vaccination rate in the model.



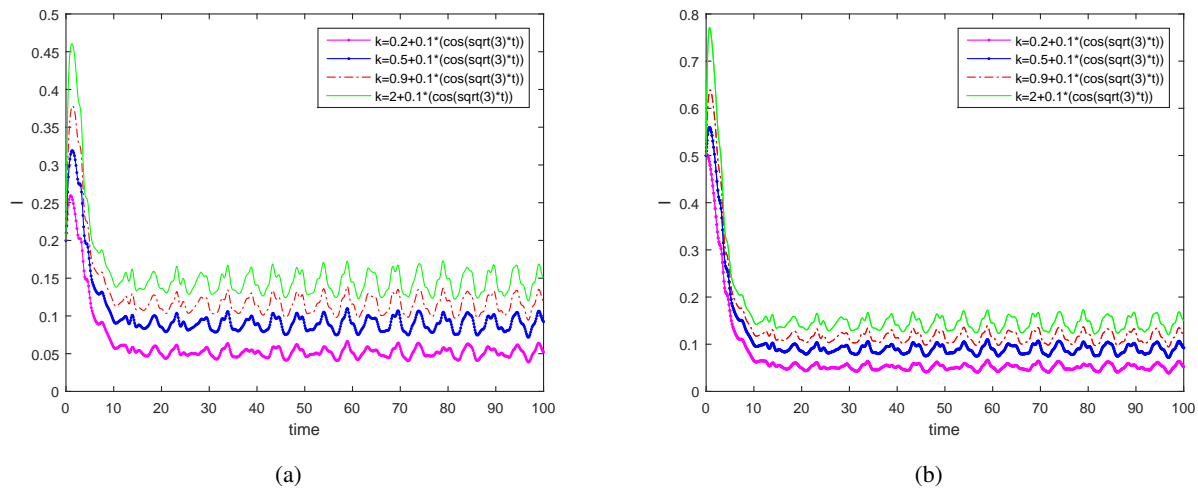
**Figure 4.** Dependence of  $I$  on the parameter  $k$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .



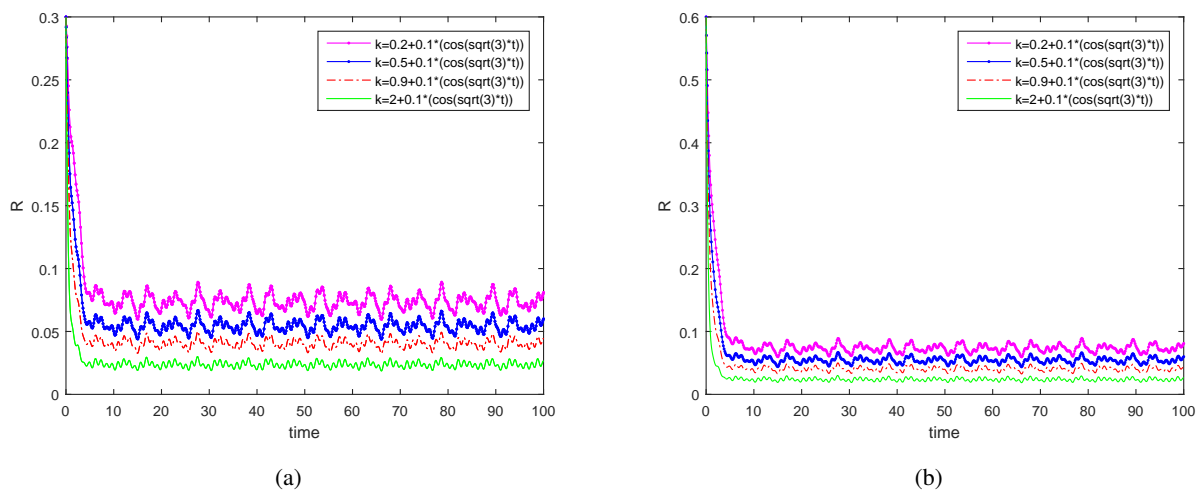
**Figure 5.** Dependence of  $R$  on the parameter  $k$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

## 7. Discussion

In this paper, we have formulated a nonautonomous SVIR epidemic model with relapse. As is known to us all, there are many disease shows seasonal behavior, taking account of seasonality in epidemic model is so important. Therefore, all the parameters in this paper are almost periodic functions. Firstly, we have proved the model (1.2) is permanence. Secondly, we have derived sufficient conditions required for existence, uniqueness and globally attractive of almost periodic solution of this system. Moreover, we have deduced that the almost periodicity of time evolution for all the populations is ensured when model parameters satisfy the conditions of Theorem 2.



**Figure 6.** Dependence of  $I$  on the parameter  $k(t)$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

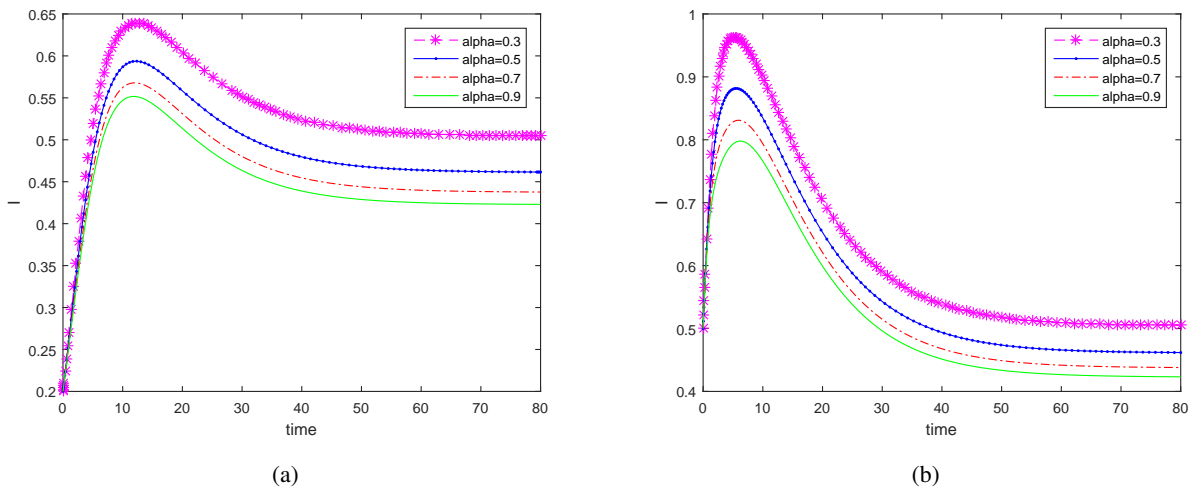


**Figure 7.** Dependence of  $R$  on the parameter  $k(t)$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

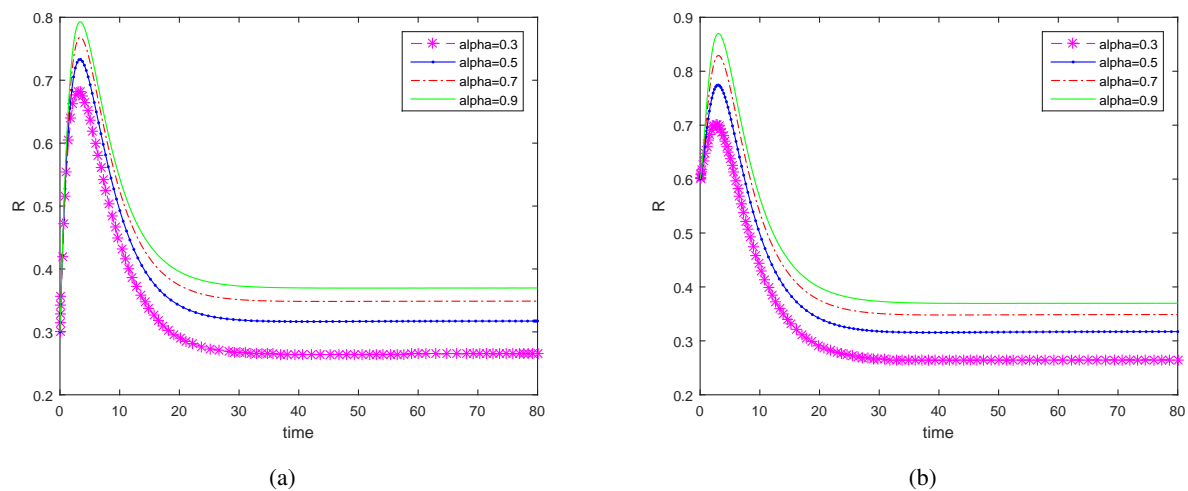
From Figures 1–3, it is easily observed that as long as the parameter satisfies the condition of Theorem 2, the equation has a globally attractive almost periodic solution, and this solution is unique.

From Figures 4–7, it is easily observed that the number of infectious people and the number of recovered people are significantly affected by  $k$ . The larger  $k$  is, the greater number of infectious people in equilibrium. This tells us that we should not only pay attention to susceptible people but also pay more attention to those who have recovered.

From Figures 9–11, it is easily observed that when the vaccination rate increases, the number of infectious people will decrease and at the same time the number of recovered patients will increase. This tells us that the government should increase publicity to strengthen people's awareness of vaccination, thereby increasing the vaccination rate.



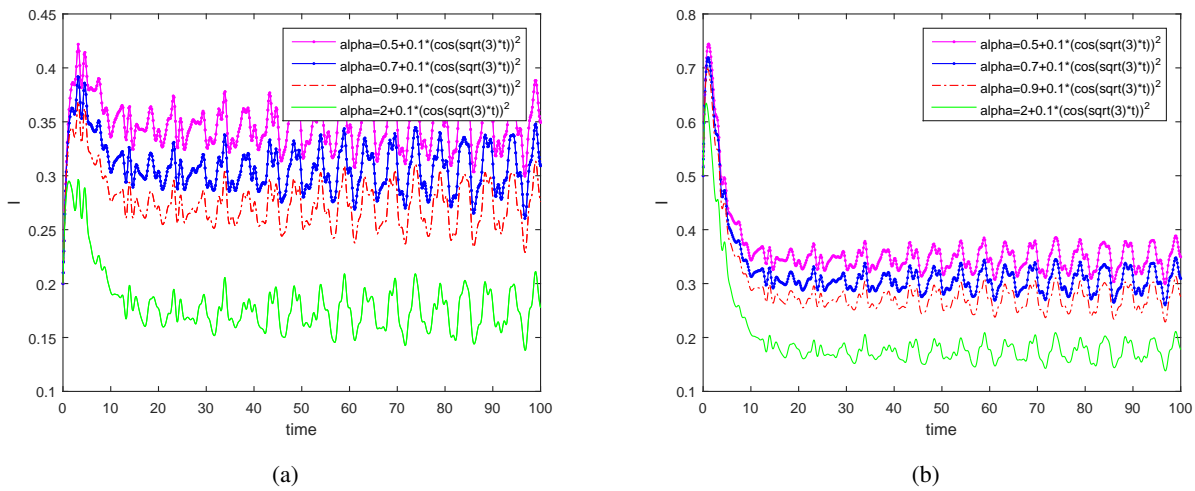
**Figure 8.** Dependence of  $I$  on the parameter  $\alpha$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .



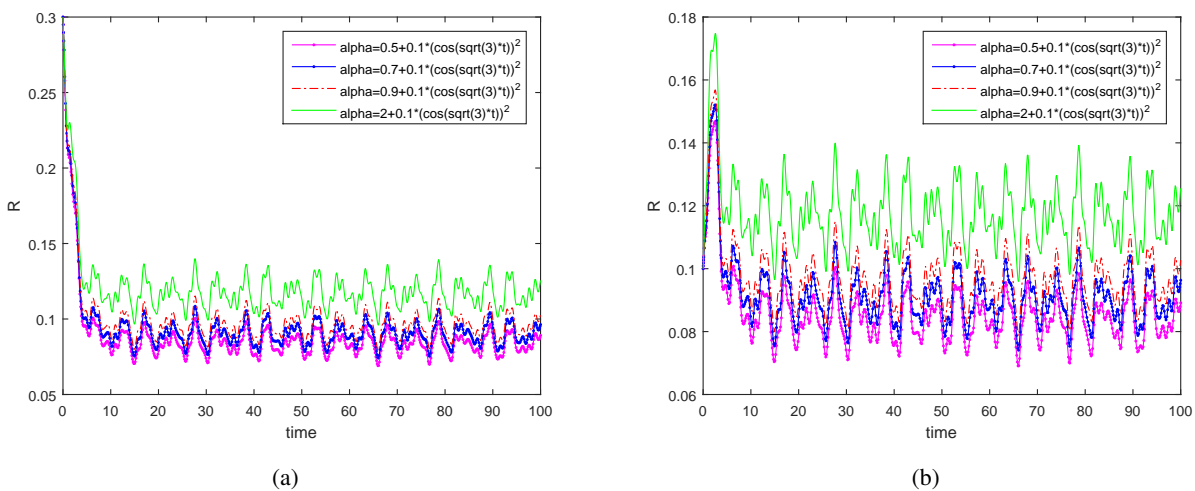
**Figure 9.** Dependence of  $R$  on the parameter  $\alpha$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

Then, we choose parameter values as follows:

$$\begin{aligned}
 \mu &= 0.2 + \frac{1}{10} \sin \sqrt{7}t, \beta = 2 - \frac{1}{10} \sin 3t, \\
 \alpha &= 0.3 + \frac{1}{2} (\cos \sqrt{2}t + \sin \sqrt{5}t)^2, \beta_1 = 0.5 + \frac{1}{10} \sin t, \\
 \gamma_1 &= 0.3 + \frac{1}{2} (\cos \sqrt{2}t)^2, \gamma = 0.1 + (\sin \sqrt{2}t)^2, \\
 k &= 0.2 + \frac{1}{10} \cos \sqrt{3}t.
 \end{aligned} \tag{7.1}$$



**Figure 10.** Dependence of  $I$  on the parameter  $\alpha(t)$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

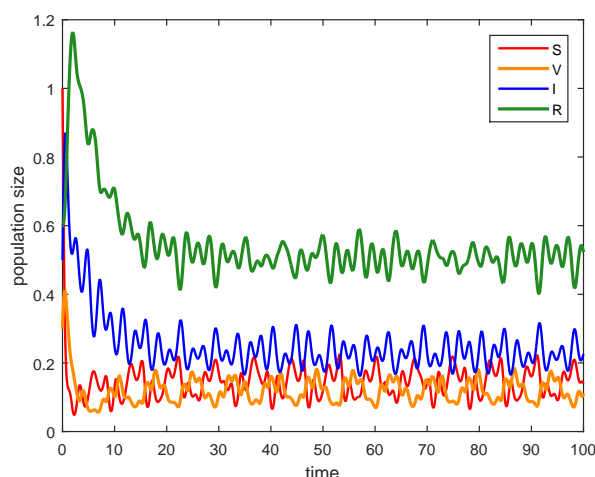


**Figure 11.** Dependence of  $R$  on the parameter  $\alpha(t)$ . (a) The initial value  $Q_0 = (0.5, 0.8, 0.2, 0.3)$ . (b) The initial value  $Q_0 = (1, 0.3, 0.5, 0.6)$ .

Obviously,  $\underline{\mu} = 0.1$ ,  $\bar{\mu} = 0.3$ ,  $\underline{\beta} = 1.9$ ,  $\bar{\beta} = 2.1$ ,  $\underline{\alpha} = 0.3$ ,  $\bar{\alpha} = 2.3$ ,  $\underline{\gamma} = 0.1$ ,  $\bar{\gamma} = 1.1$ ,  $\underline{\gamma}_1 = 0.3$ ,  $\bar{\gamma}_1 = 0.8$ ,  $\underline{\beta}_1 = 0.49$ ,  $\bar{\beta}_1 = 0.51$ ,  $\underline{k} = 0.1$ ,  $\bar{k} = 0.3$ . All the sufficient conditions given in Theorem 2 for system (1.2) are well satisfied as

$$\begin{cases} \underline{\mu} - 2\bar{\beta}M_3 = -12.5 < 0, \\ \underline{\mu} - 2\bar{\beta}_1M_3 = -2.96 < 0, \\ \underline{\mu} - 2\bar{\beta}M_1 - 2\bar{\beta}_1M_2 \approx -5.06 < 0. \end{cases}$$

Parameter values in this example fail to satisfy condition (4.1) which we have mentioned in Theorem 2. But Figure 12 shows that the model has a unique globally attractive positive almost periodic solution,



**Figure 12.** Almost periodic solution of system (1.2) with parametric values in (7.1).

which means condition (4.1) is sufficient but not necessary for Theorems 2 and 3. This problem cannot be solved at present, we shall conduct further research on this issue in the future.

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### Conflict of interest

The authors declare there is no conflict of interest.

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