

# The Dynamic Properties of a Deterministic SIR Epidemic Model in Discrete-Time

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## Abstract

In this paper, a class of discrete deterministic SIR epidemic model with vertical and horizontal transmission is studied. Based on the population assumed to be a constant size, we transform the discrete SIR epidemic model into a planar map. Then we find out its equilibrium points and eigenvalues. From discussing the influence of the coefficient parameters effected on the eigenvalues, we give the hyperbolicity of equilibrium points and determine which point is saddle, node or focus as well as their stability. Further, by deriving equations describing flows on the center manifolds, we discuss the transcritical bifurcation at the non-hyperbolic equilibrium point. Finally, we give some numerical simulation examples for illustrating the theoretical analysis and the biological explanation of our theorem.

## **Keywords**

Epidemic Model, Equilibrium Point, Transcritical Bifurcation, Center Manifold, Hyperbolicity

## **1. Introduction**

Since Kermack and McKendrick [1] proposed the Susceptible-Infective-Recovered model (or SIR for short) in 1927, a lot of glorious studies on the dynamics of the epidemic models have been presented (see [2]-[9]). The basic and important research subjects for these systems are local and global stability of the disease-free equilibrium and the endemic equilibrium, existence of periodic solutions, persistence and extinction of the disease, etc. According to the dependence on variable (*i.e.*, time), these systems were classified into two types: continuous-time system and discrete-time system.

For the epidemic models, there have been a lot of researches focusing on the case of continuous-time (see [2]-[6] and that cited therein). However, discrete-time models (or called difference equations) are also useful for

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modeling situations of epidemic. They can not only have the basic features of the corresponding continuoustime models but also provide a substantial reduction of computer time (see [10]). What is more, a lot of discretetime models are not trivial analogues of their continuous ones and simple models can even exhibit complex behavior (see [5] [10]).

In 1989, Hethcote [7] considered a class of continuous epidemic model with vertical and horizontal transmission.

$$\begin{cases} \dot{S} = -\beta SI - bS + pdI + b(S + R), \\ \dot{I} = \beta SI - dI - rI + qdI, \\ \dot{R} = rI - bR, \end{cases}$$
(1)

where *S* represents the proportion of individuals susceptible to the disease, who are born (with *b*) and die (with *d*) at the same rate b (b = d), and have mean life expectancy 1/*b*. The susceptible becomes infectious at a bilinear rate  $\beta I$ , where *I* is the proportion of infectious individuals and  $\beta$  is the contact rate. The infectious recover (*i.e.*, acquire lifelong immunity) at a rate *r*, so that 1/r is the mean infectious period. The constant *p*, *q*, 0 , <math>0 < q < 1, and p + q = 1, where *p* is the proportion of the offspring of infective parents that are susceptible individuals, and *q* is the proportion of the offspring of infective parents that are susceptible individuals, and *q* is the proportion of the offspring of infective parents that are susceptible individuals, and *q* is the proportion of the offspring of infective parents that are susceptible individuals. Because of biological meanings, a natural constraint is b - pd > 0. A similarly detailed description of the model and its dynamics may be found in [7]. In recent, Meng and Chen [8] have also studied the epidemic system (1). In their work, the basic reproductive rate determining the stability of disease-free equilibrium point and endemic equilibrium point was found out and the local and global stability of the equilibrium points have been researched by using Lyapunov function and Dulac function.

In this paper, we pay attention to the discrete situation of (1). From discussing the influence of the coefficient parameters effected on the eigenvalues, we give the hyperbolicity of equilibrium points and determine which point is saddle, node or focus as well as their stability. Further, by deriving equations describing flows on the center manifolds, we discuss the transcritical bifurcation at the non-hyperbolic equilibrium point and research how does small perturbation of coefficient parameters affect the number and stability of equilibrium points. Moreover, we give some numerical simulation examples for illustrating the theoretical analysis and explain the biological meaning of our theorem.

#### 2. Discrete SIR Epidemic Model with Vertical and Horizontal Transmission

In this section, we consider the discrete SIR epidemic model with vertical and horizontal transmission:

$$\begin{cases} S_{n+1} = -\beta S_n I_n + S_n + p dI_n + bR_n, \\ I_{n+1} = \beta S_n I_n + (1 - d - r) I_n + q dI_n, \\ R_{n+1} = r I_n + (1 - b) R_n, \end{cases}$$
(2)

where  $S_n$ ,  $I_n$  and  $R_n$  represent susceptible, infective and removed (or isolated) subgroups respectively, n represents a fixed time,  $n = 0, 1, 2, \cdots$ . It is assumed that  $S_0 > 0$ ,  $I_0 \ge 0$ ,  $R_0 \ge 0$  and  $S_0 + I_0 + R_0 = 1$ . In view of assumption that population is a constant size in [3], *i.e.*,

$$S_0 + I_0 + R_0 = 1 \tag{3}$$

system (2) can be changed into

$$\begin{cases} S_{n+1} = -\beta S_n I_n + (1-b) S_n + b(1-p) I_n + b, \\ I_{n+1} = \beta S_n I_n + (1-pb-r) I_n. \end{cases}$$
(4)

Rewrite (4) as a planar map *F*:

$$\begin{bmatrix} S\\I \end{bmatrix} \mapsto \begin{bmatrix} -\beta SI + (1-b)S + b(1-p)I + b\\\beta SI + (1-pb-r)I \end{bmatrix}.$$
(5)

It is obvious that this map has a disease-free equilibrium point P:(1,0) and an endemic equilibrium point  $Q:(S^*, I^*)$ , where  $S^* = \frac{1}{w}$ ,  $I^* = \frac{b(w-1)}{\beta + b(1-p)w}$  and  $w = \frac{\beta}{pb+r}$ .

The organization of this paper is as follows. In next section, we identify all cases of non- and hyperbolic equilibria, which is a fundament for all succeeding studies. In Section 4, we discuss the transcritical bifurcation at the disease-free equilibrium of (1), the direction and stability of the transcritical bifurcation is investigated by computing a center manifold. In Section 5, some simulations are made to demonstrate our results and the biologic explanation of the theorem is also given.

#### 3. Hyperbolic and Non-Hyperbolic Cases

In this section, we will discuss the hyperbolic and non-hyperbolic cases in a two parameters space parameter. **Theorem 1.** *The equilibrium point* P(1,0) *is non-hyperbolic if and only if* (w,b) *lies on the lines:* 

$$\ell_1: \{(w,b) | w = \beta / (2+\beta), 0 < b < 1\}$$

and

$$\ell_2: \{(w,b) | w = 1, 0 < b < 1\}$$

Otherwise, the equilibrium point P(1,0) is in one of the following types (see Table 1).

**Remark 1.** By Theorem 3.1 the domain  $\{(w,b) \in \mathbb{R}^2 | 0 < w, 0 < b < 1\}$  is divided by line  $\ell_1$  and  $\ell_2$  into three districts  $\mathcal{D}_1$ ,  $\mathcal{D}_2$  and  $\mathcal{D}_3$  for equilibrium point P (see Figure 1).

*Proof.* The Jacobian matrix of (5) at P(1,0) is

$$DF(P) = \begin{bmatrix} 1-b & -\beta-b+pb \\ 0 & 1+\beta-pb-r \end{bmatrix}$$

and its eigenvalues are

$$\lambda_1 = 1 - b, \quad \lambda_2 = 1 - \beta \left( 1 - \frac{1}{w} \right).$$

**Table 1.** Types of hyperbolic equilibrium point P(1,0).

Cases	Conditions	Eigenvalues	Properties
$\mathcal{D}_{_{1}}$	$0 < w < \beta / (2 + \beta)$	$0 < \lambda_1 < 1, \lambda_2 < -1$	Saddle
$\mathcal{D}_{_2}$	$\beta/(2+\beta) < w < 1$	$0 < \lambda_1 < 1$ , $-1 < \lambda_2 < 1$	Stable node
$\mathcal{D}_{_2}$	<i>w</i> > 1	$0<\lambda_{\scriptscriptstyle 1}<1,\lambda_{\scriptscriptstyle 2}>1$	Saddle



Figure 1. Districts for equilibrium point P.

From the assumption 0 < b < 1, we see that  $0 < \lambda_1 < 1$ . Then non-hyperbolicity will be happened in the case  $\lambda_2 = \pm 1$ . From  $\lambda_2 = -1$ , we know  $w = \beta/(2+\beta)$  which means (w,b) lies on  $\ell_1$ ; Also, From  $\lambda_2 = 1$ , we get that w = 1 and (w,b) lies on  $\ell_2$ . When  $0 < w < \beta(2+\beta)$  (referred to the case  $\mathcal{D}_1$ ), the eigenvalue  $\lambda_2$  satisfies  $\lambda_2 < -1$ , then the equilibrium point *P* is a stable node. When  $\beta(2+\beta) < w < 1$  (referred to the case  $\mathcal{D}_2$ ), the equilibrium point *P* is a node since  $-1 < \lambda_2 < 1$  and meanwhile when 1 < w, the equilibrium *P* is a saddle since  $\lambda_1 > 1$ . The proof is complete.

**Theorem 2.** There does not exist non-hyperbolic case for equilibrium point  $Q(S^*, I^*)$ . But the hyperbolicity can be divided into the following cases:

(I) When  $b < \beta$ , there exist six types for hyperbolic equilibrium point Q (see Table 2).

Where  $s_1$  and  $s_2$  satisfy  $(s_1, r) \in C_1$ :

$$\left\{ \left(s_{1}, r\right) \mid r = \left(s_{1} - b\right)^{2} / \left(4s_{1}\right), \ b_{1} = \left(\sqrt{b+1} - 1\right)^{2} \le s_{1} \le b \right\}$$

and  $(s_2, r) \in C_2$ :  $\{(s_2, r) | r = (s_2 - b)^2 / (4s_2), b < s_2 \le \beta\}$ , respectively.

(II) When  $b_1 < \beta \le b$ , there exist four types for hyperbolic equilibrium point Q (see Table 3).

Where  $s_3$  satisfies  $(s_3, r) \in C_3$ :  $\{(s_3, r) | r = (s_3 - b)^2 / (4s_3), b_1 < s_3 < \beta\}$ .

**Remark 2.** By Theorem 3.2, when  $b < \beta$ , the domain  $\{(s,r) \in \mathbb{R}^2 | -b < s, 0 < r < 1\}$  is divided by line  $\ell_3$ ,  $\mathcal{C}_1$  and  $\mathcal{C}_2$  into three districts  $\mathcal{D}_4$ ,  $\mathcal{D}_4$ ,  $\mathcal{D}_6$  and  $\mathcal{D}_7$  for equilibrium point Q (see Figure 2).

When  $b_1 < \beta \le b$ , the domain  $\{(s, r) \in \mathbb{R}^2 | -b < s, 0 < r < 1\}$  is divided by line  $\ell_4$  and  $\mathcal{C}_3$  into three districts  $\mathcal{D}_8$ ,  $\mathcal{D}_9$  and  $\mathcal{D}_9$  for equilibrium point Q (see Figure 3).

*Proof.* Performing a coordinate shift as follows:

$$\tilde{I}=I-I^*, \ \tilde{R}=R-R^*,$$

**Table 2.** Types of hyperbolic equilibrium  $Q(S^*, I^*)$ .

Cases	Conditions	Eigenvalues	Properties
$\mathcal{C}_{_{1}}$	$r = \left(s - b\right)^2 / \left(4s\right),  b_1 \le s \le b$	$-1 < \lambda_1 = \lambda_2 < 1$	Stable node
$\mathcal{C}_{_2}$	$r = \left(s - \beta\right)^2 / \left(4s\right),  b < s < \beta$	$-1 < \lambda_{_1} = \lambda_{_2} < 1$	Stable node
$\mathcal{D}_{_4}$	-b < s < 0	$\lambda_1 > 1, 0 < \lambda_2 < 1$	Saddle
$\mathcal{D}_{_{5}}$	$0 < s < s_1$	$0 < \lambda_1 < 1, 0 < \lambda_2 < 1$	Stable node
$\mathcal{D}_{_6}$	$s_1 < s < s_2$	$\lambda_{1,2}$ are complex, $\left \lambda_{1,2}\right  < 1$	Stable focus
$\mathcal{D}_7$	<i>s</i> <sub>2</sub> < <i>s</i>	$0 < \lambda_1 < 1, 0 < \lambda_2 < 1$	Stable node

**Table 3.** Types of hyperbolic equilibrium  $Q(S^*, I^*)$ .

Cases	Conditions	Eigenvalues	Properties
$\mathcal{C}_{_3}$	$r = (s-b)^2/(4s),  b \le s \le \beta$	$0 < \lambda_1 = \lambda_2 < 1$	Stable node
$\mathcal{D}_{_{8}}$	-b < s < 0	$\lambda_1 > 1,  0 < \lambda_2 < 1$	Saddle
$\mathcal{D}_{_{9}}$	$0 < s < s_{3}$	$0 < \lambda_1 < 1, 0 < \lambda_2 < 1$	Stable node
$\mathcal{D}_{_{10}}$	$s_3 < s < \beta$	$\lambda_{1,2}$ are complex, $\left \lambda_{1,2}\right  < 1$	Stable focus



and letting  $\tilde{F}$  denote the transformed F, we translate equilibrium  $Q(S^*, I^*)$  into (0,0) and discuss equilibrium point (0,0) of the map  $\tilde{F}$ . The matrix of linearization of  $\tilde{F}$  at (0,0) is

$$D\tilde{F}((0,0)) = \begin{bmatrix} 1-b-\frac{b(\beta-pb-r)}{\beta+r} & -b-r\\ \frac{b(\beta-pb-r)}{\beta+r} & 1 \end{bmatrix} = \begin{bmatrix} a-s & a-r-1\\ s & 1 \end{bmatrix},$$

where a = 1-b and  $s = \frac{b(\beta - pb - r)}{\beta + r}$ , and its eigenvalues are

$$\lambda_{1} = \frac{(1-s+a) + \sqrt{(1-s+a)^{2} - 4(a+(r-a)s)}}{2},$$

$$\lambda_{2} = \frac{(1-s+a) - \sqrt{(1-s+a)^{2} - 4(a+(r-a)s)}}{2}.$$
(6)

It is known that (0,0) is hyperbolic if and only if none of eigenvalues  $\lambda_1$ ,  $\lambda_2$  lies on the unit circle  $S^1$ . In the following we discuss the eigenvalues in two cases, *i.e.*,  $b < \beta$  and  $b_1 < \beta \le b$ . Case (I). When discriminant  $\Delta = (1 - s + a)^2 - 4(a + (r - a)s) \ge 0$ , then  $\lambda_1$  and  $\lambda_2$  are both real. Because non-

Case (I). When discriminant  $\Delta = (1 - s + a)^2 - 4(a + (r - a)s) \ge 0$ , then  $\lambda_1$  and  $\lambda_2$  are both real. Because non-hyperbolicity happens if and only if  $|\lambda_1|$  or  $|\lambda_2|$  is 1. For whether  $\lambda_1 = 1$  or  $\lambda_2 = 1$ , we can get a - r = 1. However, for positive equilibrium point Q, we have 0 < a < 1 and 0 < r < 1. Therefore, neither  $\lambda_1 = 1$  nor  $\lambda_2 = 1$  is possible. Next, let's examine  $\lambda_1 = -1$  and  $\lambda_2 = -1$ . From whether  $\lambda_1 = -1$  or  $\lambda_2 = -1$ , we can get  $\gamma = (a+1)(1-2/s)$ . By condition 0 < a < 1,  $-b < s < \beta$ , we see that (a+1)(1-2/s) > 1. This is a contradiction with r < 1, so  $\lambda_1 = -1$  and  $\lambda_2 = -1$  are impossible.

In the case of  $\Delta < 0$ ,  $\lambda_1$  and  $\lambda_2$  are a pair of conjugate complex. Since

$$|\lambda_1|^2 = |\lambda_2|^2 = \frac{1}{4} \Big[ (1-s+a)^2 + 4(a+(r-a)s) - (1-s+a)^2 \Big]$$
$$= a+(r-a)s = (s-1)b+(r-1)s+1 < 1,$$

 $\lambda_1$  and  $\lambda_2$  lie inside of  $S^1$  and the equilibrium point Q is a stable focus referred to the case  $(\mathcal{D}_6)$ . If  $\Delta = 0$ , *i.e.*,  $4rs = (1 - s - a)^2$ , the matrix has a double real eigenvalue  $\lambda_1 = \lambda_2 = (1 - s + a)/2$ . From the constraint condition  $-b < s < \beta$  of s, it is obvious that  $0 < \lambda_1 = \lambda_2 < 1$ . Therefore, equilibrium Q is stable node in the cases of  $C_1$  and  $C_2$ .

If  $\Delta > 0$ , *i.e.*  $4rs < (1-s-a)^2$  and  $-b < s < \beta$ , the eigenvalues  $\lambda_1$  and  $\lambda_2$  are different real numbers. We first discuss the case that -b < s < 0, *i.e.*  $(s,r) \in \mathcal{D}_4$ . In this case we have

$$\frac{\mathrm{d}\lambda_2}{\mathrm{d}r} = \frac{s}{\sqrt{\left(1 - s + a\right)^2 - 4\left(a + \left(r - a\right)s\right)}} < 0$$

Since

$$\lim_{r \to 0^+} \lambda_2 = \lim_{r \to 0^+} \frac{(1 - s + a) - \sqrt{(1 - s + a)^2 - 4(a + (r - a)s)}}{2} = a < 1$$

we have  $0 < \lambda_2 < 1$  for  $(s, r) \in \mathcal{D}_4$ . On the other hand, there also exists  $0 < \lambda_1 < 1$  for  $(s, r) \in \mathcal{D}_4$ . In fact, since

$$\lim_{r \to 0+} \lambda_1 = \lim_{r \to 0+} \frac{(1-s+a) + \sqrt{(1-s+a)^2 - 4(a+(r-a)s)}}{2} = 1-s > 1$$

and

$$\frac{\mathrm{d}\lambda_1}{\mathrm{d}r} = -\frac{s}{\sqrt{\left(1-s+a\right)^2 - 4\left(a+\left(r-a\right)s\right)}} > 0,$$

we have  $\lambda_1 > 1$ . Therefore, the equilibrium *Q* is a stable node as  $(s, r) \in \mathcal{D}_4$ . For the case  $0 < s < s_1$ , *i.e.*,  $(s, r) \in \mathcal{D}_5$ , we have

$$\frac{\mathrm{d}\lambda_2}{\mathrm{d}r} > 0, \quad \lim_{r \to 0^+} \lambda_2 = a > 0, \quad \lambda_2 = \frac{1 - s + a}{2} < 1,$$

and

$$\frac{\mathrm{d}\lambda_1}{\mathrm{d}\gamma} < 0, \quad \lim_{\gamma \to 0+} \lambda_1 = 1 - s < 1.$$

Therefore, the equilibrium Q is a stable node as  $(s, r) \in \mathcal{D}_5$ . Finally, we study the case of  $s_2 < s$ , *i.e.*  $(s, r) \in \mathcal{D}_7$ . We have

$$\frac{d\lambda_{1}}{dr} = -\frac{s}{\sqrt{\left(1 - s + a\right)^{2} - 4\left(a + \left(r - a\right)s\right)}} < 0, \ \lim_{r \to 0+} \lambda_{1} = a < 1, \ \lambda_{1} > \frac{1 - s + a}{2} > 0$$

Then, we have  $0 < \lambda_1 < 1$  for  $(s, r) \in \mathcal{D}_7$ . Moreover, there also has  $0 < \lambda_2 < 1$  for  $(s, r) \in \mathcal{D}_7$ . In fact that,

$$\frac{d\lambda_2}{dr} = \frac{s}{\sqrt{\left(1 - s + a\right)^2 - 4\left(a + \left(r - a\right)s\right)}} > 0, \ \lim_{r \to 0^+} \lambda_2 = 1 - s > 0, \ \lambda_2 < \frac{1 - s + a}{2} < 1$$

This means that the equilibrium Q is a stable node for  $(s, r) \in \mathcal{D}_7$ .

Case (II). When discriminant  $\Delta = (1-s+a)^2 - 4(a+(r-a)s) \ge 0$ , then  $\lambda_1$  and  $\lambda_2$  are both real. Because non-hyperbolicity happens if and only if  $|\lambda_1|$  or  $|\lambda_2|$  is 1. Similar to the proof in case (I), neither  $|\lambda_1|=1$  nor  $|\lambda_2|=1$  is possible.

When  $\Delta < 0$ ,  $\lambda_1$  and  $\lambda_2$  are a pair of conjugate complex. Since

$$|\lambda_1|^2 = |\lambda_2|^2 = a + (r-a)s < 1,$$

 $\lambda_1$  and  $\lambda_2$  lie inside of  $S^1$  and the equilibrium point Q is a stable focus referred to the case  $(\mathcal{D}_{10})$ . If  $\Delta = 0$ , the matrix has a double real eigenvalue  $\lambda_1 = \lambda_2 = (1 - s + a)/2$ . From the constraint condition  $-b < s < \beta$  of s, it is obvious that  $0 < \lambda_1 = \lambda_2 < 1$ . Therefore, equilibrium Q is stable node in the cases of  $C_3$ . If  $\Delta > 0$ , We first discuss the case that -b < s < 0, *i.e.*  $(s, r) \in \mathcal{D}_s$ . In this case we have

$$\frac{\mathrm{d}\lambda_2}{\mathrm{d}r} < 0, \quad \lim_{r \to 0^+} \lambda_2 = a < 1.$$

We know  $0 < \lambda_2 < 1$  for  $(s, r) \in \mathcal{D}_8$ . On the other hand, there exists  $0 < \lambda_1 > 1$  for  $(s, r) \in \mathcal{D}_8$ . In fact, since

$$\frac{\mathrm{d}\lambda_1}{\mathrm{d}r} > 0, \quad \lim_{r \to 0^+} \lambda_1 = 1 - s > 1$$

Therefore, the equilibrium Q is a saddle as  $(s, r) \in \mathcal{D}_8$ .

Finally, we study the case of  $0 < s < s_3$ , *i.e.*  $(s, r) \in \mathcal{D}_9$ . We easily prove  $0 < \lambda_2 < \lambda_1 < 1$  for  $(s, r) \in \mathcal{D}_9$  by same methods as in case (1). This means that the equilibrium Q is a stable node for  $(s, r) \in \mathcal{D}_9$ .

The proof is complete.

#### 4. Transcritical Bifurcation

In this section we consider the case that  $(w,b) \in \ell_2$ , where the transcritical bifurcation at equilibrium point P(1,0) will happen. The following lemmas were be derived from reference [11].

Lemma 1. ([11], Theorem 2.1.4) The map

$$\begin{cases} x \mapsto Ax + f(x, y), \\ y \mapsto Bx + g(x, y), \end{cases} \quad (x, y) \in \mathbb{R}^{c} \times \mathbb{R}^{s},$$
(7)

satisfies that A is  $c \times c$  matrix with eigenvalues of modulus one, and B is  $s \times s$  matrix with eigenvalues of modulus less than one, and

$$\begin{cases} Df(0,0) = f(0,0) = 0, \\ Dg(0,0) = g(0,0) = 0, \end{cases}$$
(8)

where f and g are  $C^r(r \ge 2)$  in some neighborhood of the origin. Then there exists a  $C^r$  center manifold for (7) which can be locally represented as a graph as follows

$$W^{c}(0) = \left\{ (x, y) \in \mathbb{R}^{c} \times \mathbb{R}^{s} \middle| y = h(x), |x| < \delta, h(0) = 0, Dh(0) = 0 \right\}$$

for  $\delta$  sufficiently small. Moreover, the dynamics of (7) restricted to the center manifold is, for *u* sufficiently small, given by the *c*-dimensional map

$$u \mapsto Au + f(u, h(u)), u \in \mathbb{R}^{c}.$$

**Lemma 2.** ([11], in page 365) A one-parameter family of  $C'(r \ge 2)$  one-dimensional maps

$$u \mapsto f(x,\mu), \ x \in \mathbb{R}^1, \mu \in \mathbb{R}^1 \tag{9}$$

having a nonhyperbolic fixed point, i.e.,

$$f(0,0) = 0, \ \frac{\partial f}{\partial x}(0,0) = 1,$$

undergoes a transcritical bifurcation at  $(x, \mu) = (0, 0)$  if

$$\frac{\partial f}{\partial \mu}(0,0) = 0, \ \frac{\partial^2 f}{\partial x \partial \mu}(0,0) \neq 0, \ \frac{\partial^2 f}{\partial x^2}(0,0) \neq 0.$$

**Theorem 3.** A transcritical bifurcation occurs at the equilibrium P when w = 1. More concretely, for w < 1 slightly there are two equilibriums: a stable point P and an unstable negative equilibrium which coalesce at w = 1 and for w > 1 slightly there are also two equilibriums: an unstable equilibrium P and a stable positive equilibrium Q. Thus an exchange of stability has occurred at w = 1.

*Proof.* For  $(w,b) \in \ell_1$ , we have  $\lambda_1 = 1$  and  $0 < \lambda_2 < 1$ . Consider w as the bifurcation parameter and write F as  $F_w$  to emphasize the dependence on w. One can easily see that the matrix  $DF_w(0,0)$  is

$$\begin{bmatrix} 1-b & -r-b \\ 0 & 1 \end{bmatrix}$$

and it has eigenvectors

$$\left(1,0\right)^{\mathrm{T}},\ \left(1,-\frac{b}{r+b}\right)^{\mathrm{T}}\tag{10}$$

corresponding to  $\lambda_1$  and  $\lambda_2$  respectively, where T means the transpose of matrices.

First, we put the matrix  $DF_w(0,0)$  into a diagonal form. Using the eigenvectors (10), we obtain the transformation

$$\begin{bmatrix} S \\ I \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & -\frac{b}{r+b} \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix}$$
(11)

with inverse

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} 1 & \frac{r+b}{b} \\ 0 & -\frac{r+b}{b} \end{bmatrix} \begin{bmatrix} S \\ I \end{bmatrix}$$
(12)

which transform system (5) into

$$\begin{bmatrix} u \\ v \end{bmatrix} \mapsto \begin{bmatrix} 1-b & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} \operatorname{red} + \begin{bmatrix} \frac{r\beta(pb+r)}{(r+b)(\beta-pb-r)}\overline{w}v - \frac{r\beta}{r+b}v^2 - \frac{r\beta}{r+b}uv + b \\ -\frac{\beta(pb+r)}{\beta-pb-r}\overline{w}v + \beta v^2 + \beta uv \end{bmatrix}.$$
(13)

Rewrite system (13) in the suspended form with assumption  $\overline{w} = w - 1 = (\beta - pb - r)/(pb + r)$ ,

$$\begin{bmatrix} u \\ v \\ \overline{w} \end{bmatrix} \mapsto \begin{bmatrix} 1-b & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ v \\ \overline{w} \end{bmatrix} + \begin{bmatrix} a_{11}v\overline{w} - a_{12}v^2 - a_{13}uv + b \\ -a_{21}v\overline{w} + a_{22}v^2 + a_{23}uv \\ 0 \end{bmatrix}$$
(14)

where  $a_{11} = \frac{r\beta(pb+r)}{(r+b)(\beta-pb-r)}$ ,  $a_{12} = a_{13} = \frac{r\beta}{r+b}$ ,  $a_{21} = \frac{\beta(pb+r)}{\beta-pb-r}$ ,  $a_{22} = a_{23} = \beta$ .

Thus, from Lemma 4.1, the stability of equilibrium (u, v) = (0, 0) near  $\overline{w} = 0$  can be determined by studying a one-parameter family of map on a center manifold which can be represented as follows,

$$W^{c}(0,0) = \{(u,v,\overline{w}) \in \mathbb{R}^{3} \mid u = h(v,\overline{w}), h(0,0) = 0, Dh(0,0) = 0\}$$

for sufficiently small v and  $\overline{w}$ .

We now want to compute the center manifold and derive the mapping on the center manifold. We assume

$$h(v,\overline{w}) = Av^2 + Bv\overline{w} + C\overline{w}^2 + \mathcal{O}(3)$$
<sup>(15)</sup>

near the origin, where  $\mathcal{O}(3)$  means terms of order  $\geq 3$ . By Lemma 4.1, those coefficients A, B and C can be determined by the equation

$$\mathcal{N}\left(h(v,\overline{w})\right) \coloneqq h\left(v + a_{21}v\overline{w} + a_{22}v^2 + a_{23}vh(v,\overline{w}),\overline{w}\right) - \left((1-b)h(v,\overline{w}) + a_{11}v\overline{w} - a_{12}v^2 - a_{13}vh(v,\overline{w}) + b\right) = 0.$$
(16)

Substituting (15) into (16) and comparing coefficients of  $v^2$ ,  $v\overline{w}$  and  $\overline{w}^2$  in (16), we get

$$\begin{cases} A + A(b-1) + a_{21} = 0\\ B + B(b-1) - a_{11} = 0\\ C + C(b-1) = 0, \end{cases}$$

from which we solve

$$A = -\frac{a_{21}}{b}, \quad B = \frac{a_{11}}{b}, \quad C = 0$$

Therefore the expression of (15) is approximately determined. Substituting (15) into (14), we obtain a one dimensional map reduced to the center manifold

$$v \mapsto \phi_{\overline{w}}(v) = v + a_{21}v\overline{w} + a_{22}v^2 - \frac{a_{21}a_{23}}{b}v^3 + \frac{a_{11}a_{23}}{b}v^2\overline{w} + \mathcal{O}(4).$$
(17)

It is easy to check that

$$\frac{\partial \phi_{\overline{w}}}{\partial \overline{w}}(0,0) = 0, \quad \frac{\partial^2 \phi_{\overline{w}}}{\partial v \partial \overline{w}}(0,0) \neq 0, \quad \frac{\partial^2 \phi_{\overline{w}}}{\partial \overline{w}^2}(0,0) \neq 0.$$
(18)

The condition (18) implies that in the study of the orbit structure near the bifurcation point terms of  $\mathcal{O}(3)$  do not qualitatively affect the nature of the bifurcation, namely they do not affect the geometry of the curves of equilibriums passing through the bifurcation point. Thus, (18) shows that the orbit structure of (17) near  $(v, \overline{w}) = (0, 0)$  is qualitatively the same as the orbit structure near  $(v, \overline{w}) = (0, 0)$  of the map

$$v \mapsto v + a_{21}v\overline{w} + a_{22}v^2. \tag{19}$$

Map (19) can be viewed as truncated normal form for the transcritical bifurcation (see Lemma 4.2). The stability of the two branches of equilibriums lying on both sides of s = 0 are easily verified.

#### 5. Simulations

In this section, we will give a simulation to illustrate the result obtained in the above section.

**Example 1.** Let p = 0.8, b = 0.1, r = 0.02 and choose three groups of initial values for  $(S_0, I_0, R_0)$  as follows:

(0.1, 0.9, 0), (0.4, 0.6, 0), (0.1, 0.3, 0.6).

If let  $\beta = 0.05$ , we see that w < 0.5 < 1 and two equilibrium points of system (2) occurs, where disease-free equilibrium point P(1,0,0) is stable and other negative equilibrium point (no meaning in biology) is unstable. The simulation result is presented in Figure 4.

If let  $\beta = 0.5$ , then w = 5 > 1 and and two equilibrium points of system (2) occurs, where disease-free equilibrium point P(1,0,0) is unstable and other positive equilibrium point Q(0.2, 0.63, 0.17) is unstable. The simulation result is presented in Figure 5.





#### 6. Biological Explanation

The conclusion in Theorem 4.1 reveals a fact that the topological structure changes at disease-free equilibrium point will take place when system (2.1) encounters small perturbation for coefficient parameters. Concretely, when parameter w = 1, system (2.1) has only one equilibrium point (1,0,0) which is structural unstable. With the change of the parameter w, the number and the stability of equilibrium point will be changed.

If let w < 1, we see that two equilibrium points of system (2) occur, where disease-free equilibrium point P(1,0,0) is stable and other negative equilibrium points (no meaning in biology) are unstable. No matter how to select the initial value, the system tends to the disease-free equilibrium point after multiple iterations. Therefore, in this case, whether selection of initial value or small perturbation for coefficient parameters, epidemic disease will tend to be eliminated, that is, the population will be in disease-free state finally.

If let w > 1, we also see that two equilibrium points of system (2) occur, where disease-free equilibrium point P(1,0,0) is unstable and other endemic disease equilibrium points  $(S^*, I^*, R^*)$  are stable. No matter how to select the initial value, the system tends to the endemic disease equilibrium point after multiple iterations. Therefore, in this case, whether selection of initial value or small perturbation for coefficient parameters, disease does not tend to be eliminated, that is, the population will be diseased state finally.

Therefore, in reality we may control the factors of contact rate, birth rate, recovery rate, etc., to achieve the aim of prevention and treatment of disease.

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