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Bifurcations and chaos in a discrete SI epidemic model with fractional order

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Abstract

In this study, a new discrete SI epidemic model is proposed and established from SI fractional-order epidemic model. The existence conditions, the stability of the equilibrium points and the occurrence of bifurcation are analyzed. By using the center manifold theorem and bifurcation theory, it is shown that the model undergoes flip and Neimark–Sacker bifurcation. The effects of step size and fractional-order parameters on the dynamics of the model are studied. The bifurcation analysis is also conducted and our numerical results are in agreement with theoretical results.

Keywords: Discrete SI epidemic model with fractional order; Center manifold theorem; Flip bifurcation; Neimark–Sacker bifurcation; Chaos

1 Introduction

Mathematical modeling plays an important role in understanding the dynamics of many infectious diseases. Thus, the use of modeling is crucial in order to analyze the spread, control strategies and the mechanisms of transmission of diseases. Over the years, numerous epidemiological models have been formulated mathematically (see, *e.g.*, [1-6]). Although most of these models have been restricted to integer-order differential equations (IDEs), in the last three decades, it has turned out that many problems in different fields such as sciences, engineering, finance, economics and in particular epidemiology can be described successfully by the fractional-order differential equations (FDEs) (see, e.g., [7-12]). A property of these fractional-order models is their nonlocal property which does not exist in IDEs. Nonlocal property means that the next state of a model depends not only upon its current state but also upon all of its historical states [13]. The transformation of an integer-order model into a fractional-order model needs to be precise as to the order of differentiation α : a small change in α may cause a big change in the final results [14]. FDEs can be used to model certain phenomena which cannot be adequately modeled by IDEs [15]. FDEs are commonly used on biological systems since they are related in a natural way to systems with memory. Similar to a nonlinear differential system, a nonlinear fractional differential system may also have complex dynamics, such as chaos and bifurcation. Studying chaos in fractional-order dynamical systems is an attractive and interesting topic [16].

Discrete models constructed by the discretization of continuous models have been used to describe some epidemic models. Many significant and meaningful types of such research can be found in [17-27] and the references therein. The reason for the use of dis-



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crete models is that statistical data on epidemics are collected in discrete times and hence comparing data with output of a discrete model may be easier [28]. In general, the dynamical behavior of discrete fractional-order epidemic models may exhibit phenomena such as the period-doubling and chaotic behavior. Iwami *et al.* [29] have proposed a mathematical model in a continuous-time version to interpret a model of the spread of avian–human influenza epidemic. In [29], the dynamics of the spread into bird population and between bird and human populations have been thoroughly studied. The main objective of the current study is to extend the research done in [29] by employing fractional models with discrete-time systems. In this paper, a discrete-time SI is established by SI fractional-order epidemic model. This model contains two parameters in addition to those already existing in the original SI model proposed in [29]; time step parameter and fractional-order parameter. The occurrence of bifurcations as these parameters are varied is shown in Section 7. The stability of fixed points, the emergence of flip bifurcation and Neimark–Sacker bifurcation are also studied. Using the discretized-time SI model with fractional order is a new topic, thus this paper provides a new contribution to the literature.

This paper is organized as follows. In Section 2, the fractional-time SI epidemic model from its continuous-time counterpart is constructed. In Section 3, the discretized-time SI model with fractional-order parameter is established. Section 4 discusses the existence and stability of fixed points of the discretized-time SI model with fractional order. In Sections 5 and 6, flip and Neimark–Sacker bifurcations are demonstrated, respectively. In Section 7, some numerical examples are shown. A brief discussion of our results is given in Section 8.

2 The FDE epidemic model

Let us consider the following *IDE* epidemic model:

$$\frac{dS(t)}{dt} = \Lambda - \mu S - \beta SI,$$

$$\frac{dI(t)}{dt} = \beta SI - (\mu + r)I.$$
(2.1)

This model is constructed by Iwami *et al.* [29] to explain the spreads of avian influenza through the bird world and describes the interactions between them. The population in this model is divided into susceptible birds with size *S* and infected birds with size *I*. The new birds birth rate is expressed by the parameter Λ . Susceptible birds die at the rate μ and infected birds die at the rate $\mu + r$, where *r* is the additional death rate mediated by avian influenza. The parameter β is the bilinear incidence rate.

There are several definitions of fractional derivatives [30, 31]. One of the most common definitions is the Caputo definition [32]. This definition is often used in real applications and shown in Definition 1.

Definition 1 The fractional integral of order $\beta \in \mathbb{R}^+$ of the function f(t), t > 0, is defined by

$$I^{\beta}f(t) = \int_0^t \frac{(t-s)^{\beta-1}}{\Gamma(\beta)} f(s) \, ds,$$

and the fractional derivative of order $\alpha \in (n - 1, n)$ of f(t), t > 0, is defined by

$$D^{\alpha}f(t) = I^{n-\alpha}D^n f(t), \quad \alpha > 0,$$

where $f^{(n)}$ represents the *n*th order derivative of f(t), $n = [\alpha]$ is the value of α rounded up to the nearest integer, I^{β} is the β th order Riemann-Liouville integer operator and $\Gamma(\cdot)$ is Euler's Gamma function. The operator D^{α} is called the ' α th order Caputo differential operator'.

Now, the fractional-order form of the SI epidemic model (2.1) can be formulated as follows:

$$D_t^{\alpha} S(t) = \Lambda - \mu S - \beta SI,$$

$$D_t^{\alpha} I(t) = \beta SI - (\mu + r)I,$$
(2.2)

where D_t^{α} represents the Caputo fractional derivative, t > 0, and α is the fractional order satisfying $\alpha \in (0, 1]$.

3 Discretization process

There are many discretization methods that have been used to construct the discretetime model using continuous-time methods such as explicit and implicit Euler's method, Runge–Kutta method, predictor-corrector method and nonstandard finite difference methods [33–37]. Some of them are approximation for the derivative and some for the integral. In [38, 39] a discretization process was introduced to discretize FDEs. This discretization method is an approximation for the right hand side of the differential equation has the formula $D^{\alpha}x(t) = f(x(t)), t > 0, \alpha \in (0, 1)$. This method is now applied to the fractional-order SI model (2.2). Assume that $S(0) = S_0$ and $I(0) = I_0$ are the initial conditions of system (2.2). So, the discretization of the system (2.2) is given by the following formulas:

$$D^{\alpha}S(t) = \Lambda - \mu S\left(\left[\frac{t}{x}\right]x\right) - \beta S\left(\left[\frac{t}{x}\right]x\right)I\left(\left[\frac{t}{x}\right]x\right),$$

$$D^{\alpha}I(t) = \beta S\left(\left[\frac{t}{x}\right]x\right)I\left(\left[\frac{t}{x}\right]x\right) - (\mu + r)I\left(\left[\frac{t}{x}\right]x\right).$$

(3.1)

First, let $t \in [0, h)$, $t/h \in [0, 1)$. Then

$$D_{t}^{\alpha}S(t) = \Lambda - \mu S_{0} - \beta S_{0}I_{0},$$

$$D_{t}^{\alpha}I(t) = \beta S_{0}I_{0} - (\mu + r)I_{0},$$
(3.2)

and the solution of (3.2) reduces to

$$S_{1}(t) = S_{0} + J^{\alpha} [\Lambda - \mu S_{0} - \beta S_{0}I_{0}] = S_{0} + \frac{t^{\alpha}}{\Gamma(1+\alpha)} [\Lambda - \mu S_{0} - \beta S_{0}I_{0}],$$

$$I_{1}(t) = I_{0} + J^{\alpha} [\beta S_{0}I_{0} - (\mu + r)I_{0}] = I_{0} + \frac{t^{\alpha}}{\Gamma(1+\alpha)} [\beta S_{0}I_{0} - (\mu + r)I_{0}].$$
(3.3)

Second, let $t \in [h, 2h)$, which makes $1 \le t/h < 2$. Thus, we obtain

$$D_{t}^{\alpha}S(t) = \Lambda - \mu S_{1} - \beta S_{1}I_{1},$$

$$D_{t}^{\alpha}I(t) = \beta S_{1}I_{1} - (\mu + r)I_{1},$$
(3.4)

which have the following solutions:

$$S_{2}(t) = S_{1}(x) + \frac{(t-h)^{\alpha}}{\Gamma(1+\alpha)} \Big[\Lambda - \mu S_{1}(x) - \beta S_{1}(x) I_{1}(x) \Big],$$

$$I_{2}(t) = I_{1}(x) + \frac{(t-h)^{\alpha}}{\Gamma(1+\alpha)} \Big[\beta S_{1}(x) I_{1}(x) - (\mu+r) I_{1}(x) \Big].$$
(3.5)

Repeating the discretization process *n* times yields

$$S_{n+1}(t) = S_n(nx) + \frac{(t-nh)^{\alpha}}{\Gamma(1+\alpha)} \Big[\Lambda - \mu S_n(nx) - \beta S_n(nx) I_n(nx) \Big],$$

$$I_{n+1}(t) = I_n(nx) + \frac{(t-nh)^{\alpha}}{\Gamma(1+\alpha)} \Big[\beta S_n(nx) I_n(nx) - (\mu+r) I_n(nx) \Big],$$
(3.6)

where $t \in [nh, (n + 1)h)$. For $t \rightarrow (n + 1)h$, system (3.6) is reduced to

$$S_{n+1} = S_n + \frac{h^{\alpha}}{\Gamma(1+\alpha)} [\Lambda - \mu S_n - \beta S_n I_n],$$

$$I_{n+1} = I_n + \frac{h^{\alpha}}{\Gamma(1+\alpha)} [\beta S_n I_n - (\mu+r) I_n].$$
(3.7)

Remark 3.1 It should be noticed that if $\alpha \rightarrow 1$ in (3.7), the Euler discretization of SI model is obtained.

4 Stability of fixed points

In this section, an approach as in [40] is employed. The stability of the system (3.7) is studied around its fixed points. It is clear that the model (3.7) has always a disease free equilibrium point $E_0 = (\frac{\Lambda}{\mu}, 0)$ and an endemic equilibrium point $E_1 = (\frac{\mu+r}{\beta}, \frac{\Lambda}{\mu+r} - \frac{\mu}{\beta})$. Furthermore, the system (3.7) has basic reproduction number $\Re_0 = \frac{\beta\Lambda}{\mu(\mu+r)} > 1$, then E_1 can be reformulated as $E_1 = (\frac{\mu+r}{\beta}, \frac{\mu(\Re_0-1)}{\beta})$. It can be seen that the free equilibrium point E_0 always exists while the positive endemic E_1 exists only when $\Re_0 > 1$. So as to analyze the dynamical properties of (3.7), we compute the Jacobian matrix *J* of (3.7), and we evaluate at the fixed point $E = (S^*, I^*)$

$$J(S^*, I^*) = \begin{pmatrix} 1 - \frac{h^{\alpha}}{\Gamma(1+\alpha)}(\mu + \beta I^*) & -\frac{h^{\alpha}}{\Gamma(1+\alpha)}\beta S^* \\ \frac{h^{\alpha}}{\Gamma(1+\alpha)}\beta I^* & 1 + \frac{h^{\alpha}}{\Gamma(1+\alpha)}(\beta S^* - \mu - r) \end{pmatrix}.$$
(4.1)

In order to study the stability of the fixed points of the system (3.7), the two following lemmas are employed.

Lemma 4.1 ([41, 42]) Let λ_1 and λ_2 be the two roots of matrix J(M), we have the following definitions:

- (i) If $|\lambda_1| < 1$ and $|\lambda_2| < 1$, then the equilibrium point of $M(x^*, y^*)$ is locally asymptotically stable (sink).
- (ii) If $|\lambda_1| > 1$ and $|\lambda_2| > 1$, then the equilibrium point of $M(x^*, y^*)$ is unstable (source).
- (iii) If $|\lambda_1| < 1$ and $|\lambda_2| > 1$ (or $|\lambda_1| > 1$ and $|\lambda_2| < 1$), then the equilibrium point of $M(x^*, y^*)$ is unstable (saddle).
- (iv) If $|\lambda_1| = 1$ or $|\lambda_2| = 1$, then the equilibrium point of $M(x^*, y^*)$ is called non-hyperbolic.

Lemma 4.2 ([43, 44]) Let $F(\lambda) = \lambda^2 - \operatorname{Tr} \lambda + \operatorname{Det}$. Suppose that $F(1) > 0, \lambda_1, \lambda_2$ are the two *roots of* $F(\lambda) = 0$ *. Then*

- (i) $|\lambda_1| < 1$ and $|\lambda_2| < 1$ if and only if F(-1) > 0 and Det < 1;
- (ii) $|\lambda_1| < 1$ and $|\lambda_2| > 1$ or $(|\lambda_1| > 1$ and $|\lambda_2| < 1)$ if and only if F(-1) < 0;
- (iii) $|\lambda_1| > 1$ and $|\lambda_2| > 1$ if and only if F(-1) > 0 and Det > 1;
- (iv) $\lambda_1 = -1$ and $\lambda_2 \neq 1$ if and only if F(-1) = 0 and Tr $\neq 0, 2$;
- (v) λ_1 and λ_2 are complex and $|\lambda_1| = |\lambda_2|$ if and only if $\operatorname{Tr}^2 4\operatorname{Det} < 0$ and $\operatorname{Det} = 1$.

Based on Lemmas 4.1 and 4.2, the following results can be achieved.

Theorem 4.3 If $\mathfrak{R}_0 < 1$, then the free-equilibrium point E_0 has at least four different topological types for all its values of parameters

- (i) E_0 is a sink if $0 < h < \min\{\sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{\mu}}, \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{(\mu+r)(1-\Re_0)}}\}$. (ii) E_0 is a source if $h > \max\{\sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{\mu}}, \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{(\mu+r)(1-\Re_0)}}\}$. (iii) E_0 is a saddle if $\sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{(\mu+r)(1-\Re_0)}} < h < \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{\mu}}$ or $\sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{\mu}} < h < \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{(\mu+r)(1-\Re_0)}}$. (iv) E_0 is non-hyperbolic if $h = \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{\mu}}$ or $h = \sqrt[\alpha]{\frac{2\Gamma(1+\alpha)}{(\mu+r)(1-\Re_0)}}$.

Proof The Jacobian matrix of E_0 is

$$J(E_0) = \begin{pmatrix} 1 - \frac{h^{\alpha}\mu}{\Gamma(1+\alpha)} & \frac{-h^{\alpha}\beta\Lambda}{\mu\Gamma(1+\alpha)} \\ 0 & 1 - \frac{h^{\alpha}(\mu+r)(1-\mathfrak{N}_0)}{\Gamma(1+\alpha)} \end{pmatrix}.$$

The eigenvalues of $J(E_0)$ are $\lambda_1 = 1 - \frac{h^{\alpha}\mu}{\Gamma(1+\alpha)}$ and $\lambda_2 = 1 - \frac{h^{\alpha}(\mu+r)(1-\mathfrak{N}_0)}{\Gamma(1+\alpha)}$ where $0 < \alpha \leq 1$ and *h*, $\frac{h^{\alpha}}{\Gamma(1+\alpha)} > 0$. Hence applying the stability conditions using Lemma 4.1 the results (i)–(iv) can be achieved.

Theorem 4.4 If $\Re_0 > 1$, we have

- (i) E_1 is asymptotically stable (sink) if one of the following conditions holds:
 - (i.1) $\Delta \ge 0$ and $0 < h < h_1$.
 - (i.2) $\Delta < 0$ and $0 < h < h_2$.
- (ii) E_1 is unstable (source) if one of the following conditions holds:
 - (ii.1) $\Delta \ge 0$ and $h > h_3$.
 - (ii.2) $\Delta < 0$ and $h > h_2$.
- (iii) E_1 is unstable (saddle) if $\Delta \ge 0$ and $h_1 < h < h_3$.
- (iv) E_1 is non-hyperbolic if one of the following conditions holds:
 - (iv.1) $\Delta \geq 0$ and $h = h_1$ or h_3 ,
 - (iv.2) $\Delta < 0$ and $h = h_2$,

where

$$\begin{split} h_1 &= \sqrt[\alpha]{\frac{(\mu\mathfrak{R}_0 - \sqrt{\Delta})\Gamma(1+\alpha)}{\mu(\mu+r)(\mathfrak{R}_0 - 1)}},\\ h_2 &= \sqrt[\alpha]{\frac{\mathfrak{R}_0\Gamma(1+\alpha)}{(\mu+r)(\mathfrak{R}_0 - 1)}},\\ h_3 &= \sqrt[\alpha]{\frac{(\mu\mathfrak{R}_0 + \sqrt{\Delta})\Gamma(1+\alpha)}{\mu(\mu+r)(\mathfrak{R}_0 - 1)}}, \end{split}$$

and

$$\Delta = \left[\mu(\mathfrak{R}_0 - 2)\right]^2 - 4\mu r(\mathfrak{R}_0 - 1).$$

Proof The Jacobian matrix of E_1 can be written as

$$J(E_1) = \begin{pmatrix} 1 - \frac{h^{\alpha} \mu \Re_0}{\Gamma(1+\alpha)} & \frac{-h^{\alpha}(\mu+r)}{\Gamma(1+\alpha)} \\ \frac{h^{\alpha} \mu(\Re_0 - 1)}{\Gamma(1+\alpha)} & 1 \end{pmatrix}.$$

The characteristic equation of $J(E_1)$ has the form

$$\lambda^2 - \operatorname{Tr}(J(E_1))\lambda + \operatorname{Det}(J(E_1)) = 0,$$

where

$$\operatorname{Tr}(J(E_1)) = 2 - \frac{h^{\alpha} \mu \mathfrak{N}_0}{\Gamma(1+\alpha)}$$

and

$$\operatorname{Det}(J(E_1)) = \frac{h^{2\alpha}\mu(\mu+r)(\mathfrak{R}_0-1)}{\Gamma^2(1+\alpha)} - \frac{h^{\alpha}\mu\mathfrak{R}_0}{\Gamma(1+\alpha)} + 1.$$

Then the characteristic equation $J(E_1)$ has two eigenvalues, which are

$$\lambda_{1,2} = 1 - \frac{h^{\alpha} \mu \Re_0}{2\Gamma(1+\alpha)} \pm \frac{h^{\alpha} \sqrt{\Delta}}{2\Gamma(1+\alpha)}$$

By applying Lemmas 4.1, 4.2 and the Jury conditions [45], the stability conditions (i)–(iv) can be achieved. $\hfill \Box$

From the above analysis if the statement (iv.1) of Theorem 4.4 holds, then one of the eigenvalues of $J(E_1)$ is -1 and the other is neither 1 nor -1. The statement (iv.1) can be reformulated as follows:

$$(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_1 \cup \Omega_2,$$

where

$$\Omega_1 = \{(\alpha, h, \beta, \Lambda, \mu, r) : h = h_1, \Delta \ge 0, \mathfrak{R}_0 > 1\},\$$

$$\Omega_2 = \{(\alpha, h, \beta, \Lambda, \mu, r) : h = h_3, \Delta \ge 0, \mathfrak{R}_0 > 1\}.$$

If the parameter *h* varies in the neighborhood of h_1 and $(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_1$ or h_3 and $(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_2$, the system (3.7) may undergo a flip bifurcation of equilibrium E_1 .

When the statement (iv.2) of Theorem 4.4 holds, then the two eigenvalues of $J(E_1)$ are a pair of conjugate complex numbers and the modules of each of them equals 1. The statement (iv.2) can be reformulated as follows:

$$(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_3$$
,

where

$$\Omega_3 = \{(\alpha, h, \beta, \Lambda, \mu, r) : h = h_2, \Delta < 0, \mathfrak{R}_0 > 1\}.$$

If the parameter *h* varies in the neighborhood of h_2 and $(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_3$, the system (3.7) may undergo a Neimark–Sacker bifurcation of equilibrium E_1 .

5 Flip bifurcation analysis

Flip bifurcation of the equilibrium point E_1 when parameters $(\alpha, h, \beta, \Lambda, \mu, \gamma)$ vary in the small neighborhood of Ω_1 or Ω_2 is discussed in this section. Let $A_1 = \frac{h^{\alpha}}{\Gamma(1+\alpha)}$ and A_* be a perturbation of bifurcation parameter, then a perturbed form of model (3.7) can be formulated as follows:

$$S_{n+1} = S_n + (A_1 + A_*)[\Lambda - \mu S_n - \beta S_n I_n],$$

$$I_{n+1} = I_n + (A_1 + A_*)[\beta S_n I_n - (\mu + r)I_n].$$
(5.1)

We translate $E_1(S^*, I^*)$ to the origin by using transformations $X_n = S_n - S^*$ and $Y_n = I_n - I^*$. Then (5.1) can reformulated as follows:

$$\begin{pmatrix} X_{n+1} \\ Y_{n+1} \end{pmatrix} = \begin{pmatrix} a_{11}X_n + a_{12}Y_n + a_{13}X_nY_n + (b_{11}X_n + b_{12}Y_n + b_{13}X_nY_n)A_* \\ a_{21}X_n + a_{22}Y_n + a_{23}X_nY_n + (b_{21}X_n + b_{22}Y_n + b_{23}X_nY_n)A_* \end{pmatrix},$$
(5.2)

where

$$\begin{aligned} a_{11} &= 1 - \mu \Re_0 A_1, \qquad a_{12} = -(\mu + r)A_1, \qquad a_{13} = -\beta A_1, \\ b_{11} &= -\mu \Re_0, \qquad b_{12} = -(\mu + r), \qquad b_{13} = -\beta, \\ a_{21} &= \mu(\Re_0 - 1)A_1, \qquad a_{22} = 1, \qquad a_{23} = \beta A_1, \\ b_{21} &= \mu(\Re_0 - 1), \qquad b_{22} = 0, \qquad b_{23} = \beta, \end{aligned}$$

and $A = A_1$.

Let $T_1 = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$, then the generalized eigenvectors of T_1 corresponding to the eigenvalues λ_1 and λ_2 where $\lambda_1 = -1$ and $|\lambda_2| \neq 1$ are $\binom{K_1}{K_2} = K_2 \begin{pmatrix} \frac{-a_{12}}{1+a_{11}} \\ 1 \end{pmatrix}$ and $\binom{K_3}{K_4} = K_4 \begin{pmatrix} \frac{-a_{12}}{a_{11}-\lambda_2} \\ 1 \end{pmatrix}$,

respectively. Here, we choose $K_2 = -(1 + a_{11})$ and $K_4 = -(a_{11} - \lambda_2)$. Then we have an invertible matrix $T_2 = \binom{K_1 \ K_3}{K_2 \ K_4} = \binom{a_{12} \ a_{12}}{-1 - a_{11} \ \lambda_2 - a_{11}}$. Consider the following transformation:

$$\begin{pmatrix} X_n \\ Y_n \end{pmatrix} = T_2 \begin{pmatrix} u_n \\ \upsilon_n \end{pmatrix}.$$

Taking T_2^{-1} on both sides of (5.2), we obtain

$$\begin{pmatrix} u_{n+1} \\ v_{n+1} \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} u_n \\ v_n \end{pmatrix} + \begin{pmatrix} f_1(X_n, Y_n, A_*) \\ f_2(X_n, Y_n, A_*) \end{pmatrix},$$
(5.3)

where

$$f_{1}(X_{n}, Y_{n}, A_{*}) = \frac{(\lambda_{2} - a_{11})}{a_{12}(\lambda_{2} + 1)} (a_{13}X_{n}Y_{n} + (b_{11}X_{n} + b_{12}Y_{n} + b_{13}X_{n}Y_{n})A_{*})$$

$$- \frac{1}{(\lambda_{2} + 1)} (a_{23}X_{n}Y_{n} + (b_{21}X_{n} + b_{22}Y_{n} + b_{23}X_{n}Y_{n})A_{*}),$$

$$f_{2}(X_{n}, Y_{n}, A_{*}) = \frac{(a_{11} + 1)}{a_{12}(\lambda_{2} + 1)} (a_{13}X_{n}Y_{n} + (b_{11}X_{n} + b_{12}Y_{n} + b_{13}X_{n}Y_{n})A_{*}),$$

$$+ \frac{1}{(\lambda_{2} + 1)} (a_{23}X_{n}Y_{n} + (b_{21}X_{n} + b_{22}Y_{n} + b_{23}X_{n}Y_{n})A_{*}),$$

$$X_{n} = a_{12}(u_{n} + v_{n}), \qquad Y_{n} = -(1 + a_{11})u_{n} + (\lambda_{2} - a_{11})v_{n},$$

and

$$X_n Y_n = -a_{12}(1+a_{11})u_n^2 + a_{12}(\lambda_2 - 2a_{11} - 1)u_n \upsilon_n + a_{12}(\lambda_2 - a_{11})\upsilon_n^2$$

Now, the center manifold $W^c(0,0,0)$ of (5.3) at the fixed point (0,0) in a small neighborhood of $A_* = 0$ can be formulated. Hence, based on the center manifold theorem, we know there exists a center manifold

$$W^{c}(0,0,0) = \left\{ (u_{n}, v_{n}, A_{*}) \in \mathbb{R}^{3} : v_{n} = \varphi(u_{n}, A_{*}), \varphi(0,0) = 0, D\varphi(0,0) = 0 \right\},\$$

for u_n , A_* sufficiently small. We assume a center manifold of the form

$$\varphi(u_n, A_*) = a_1 u_n^2 + a_2 A_* u_n + 0((|u_n| + |A_*|)^3),$$

which must satisfy

$$N(\varphi(u_n, A_*)) = \varphi(-u_n + f_1(u_n, \varphi(u_n, A_*)), A_*) - \lambda_2 \varphi(u_n, A_*) - f_2(u_n, \varphi(u_n, A_*), A_*)$$

= 0. (5.4)

By equating coefficients of like powers in (5.4) to zero, we obtain

$$\begin{split} a_1 &= \frac{(1+a_{11})[a_{13}(1+a_{11})+a_{12}a_{23}]}{\lambda_2^2-1}, \\ a_2 &= \frac{(1+a_{11})[b_{12}(1+a_{11})-a_{12}b_{11}]}{a_{12}(\lambda_2+1)^2} + \frac{b_{22}(1+a_{11})-a_{12}b_{21}}{(\lambda_2+1)^2}. \end{split}$$

Therefore, the map f, which is model (3.7) restricted to the center manifold $W^c(0, 0, 0)$ takes the form

$$f(u_n) = -u_n + \varphi_1 u_n A_* + \varphi_2 u_n A_*^2 + \varphi_3 u_n^2 + \varphi_4 u_n^2 A_* + \varphi_5 u_n^3 + 0((|u_n| + |A_*|)^4),$$
(5.5)

where

$$\begin{split} \varphi_{1} &= \frac{1}{a_{12}(\lambda_{2}+1)} \Big\{ a_{12} \Big[b_{11}(\lambda_{2}-a_{11}) - a_{12}b_{21} \Big] - (1+a_{11}) \Big[b_{12}(\lambda_{2}-a_{11}) - a_{12}b_{22} \Big] \Big\}, \\ \varphi_{2} &= \frac{a_{2}}{a_{12}(\lambda_{2}+1)} \Big\{ a_{12} \Big[b_{11}(\lambda_{2}-a_{11}) - a_{12}b_{21} \Big] + (\lambda_{2}-a_{11}) \Big[b_{12}(\lambda_{2}-a_{11}) - a_{12}b_{22} \Big] \Big\}, \\ \varphi_{3} &= \frac{-1}{\lambda_{2}+1} \Big\{ (1+a_{11}) \Big[a_{13}(\lambda_{2}-a_{11}) - a_{12}a_{23} \Big] \Big\}, \\ \varphi_{4} &= \frac{1}{a_{12}(\lambda_{2}+1)} \Big\{ a_{2}a_{12}(\lambda_{2}+1) \Big[a_{13}(\lambda_{2}-a_{11}) - a_{12}a_{23} \Big] \\ &\quad + a_{1}a_{12} \Big[b_{11}(\lambda_{2}-a_{11}) - a_{12}b_{21} \Big] \\ &\quad + a_{1}(\lambda_{2}-a_{11}) \Big[b_{12}(\lambda_{2}-a_{11}) - a_{12}b_{22} \Big] \\ &\quad - a_{12}(1+a_{11}) \Big[b_{13}(\lambda_{2}-a_{11}) - a_{12}b_{23} \Big] \Big\}, \end{split}$$
(5.6)

According to the flip bifurcation, the discriminatory quantities χ_1 and χ_1 are given by

$$\chi_{1} = \left(\frac{\partial^{2} f}{\partial u_{n} \partial A_{*}} + \frac{1}{2} \frac{\partial f}{\partial A_{*}} \frac{\partial^{2} f}{\partial u_{n}^{2}} \right) \Big|_{(0,0)},$$

$$\chi_{2} = \left(\frac{1}{6} \frac{\partial^{3} f}{\partial u_{n}^{3}} + \left(\frac{1}{2} \frac{\partial^{2} f}{\partial u_{n}^{2}} \right)^{2} \right) \Big|_{(0,0)}.$$
(5.7)

Thus, $\chi_1 = \varphi_1$ and $\chi_2 = \varphi_5 + \varphi_3^2$. Therefore according to flip bifurcation conditions in [46], the following theorem can be stated.

Theorem 5.1 If $\chi_2 \neq 0$, and the parameter A_* alters in the limiting region of the point (0,0), then the system (5.1) passes through a flip bifurcation at the point $E_1(S^*, I^*)$. Further, the period-2 points that bifurcate from the fixed point $E_1(S^*, I^*)$ are stable if $\chi_2 > 0$ and unstable if $\chi_2 < 0$.

6 Neimark-Sacker bifurcation

A Neimark–Sacker bifurcation of the equilibrium point E_1 occurs when parameters $(\alpha, h, \beta, \Lambda, \mu, \gamma)$ vary in the small neighborhood of Ω_3 . A perturbation form of model (3.7) can be written as follows:

$$S_{n+1} = S_n + (A_2 + A^*) [\Lambda - \mu S_n - \beta S_n I_n],$$

$$I_{n+1} = I_n + (A_2 + A^*) [\beta S_n I_n - (\mu + r) I_n],$$
(6.1)

where $|A^*| \ll 1$ is a limited perturbation parameter. Let $X_n = S_n - S^*$, $Y_n = I_n - I^*$, then the fixed point $E_1(S^*, I^*)$ to (0,0) can be retranslated and (6.1) can be reformulated as follows:

$$\begin{pmatrix} X_{n+1} \\ Y_{n+1} \end{pmatrix} = \begin{pmatrix} c_{11}X_n + c_{12}Y_n + c_{13}X_nY_n \\ c_{21}X_n + c_{22}Y_n + c_{23}X_nY_n \end{pmatrix},$$
(6.2)

where

$$\begin{aligned} c_{11} &= 1 - \mu \Re_0 (A_2 + A^*), \qquad c_{12} = -(\mu + r) (A_2 + A^*), \qquad c_{13} = -\beta (A_2 + A^*), \\ c_{21} &= \mu (\Re_0 - 1) (A_2 + A^*), \qquad c_{22} = 1, \qquad c_{23} = \beta (A_2 + A^*), \end{aligned}$$

and $A = A_2$.

The characteristic equation associated with the linearization system of model (6.2) at (0,0) is

$$\omega^{2} + p(A^{*})\omega + q(A^{*}) = 0, \tag{6.3}$$

where

$$p(A^*) = -2 - G(A_2 + A^*),$$

$$q(A^*) = 1 + G(A_2 + A^*) + H(A_2 + A^*)^2,$$

$$G = c_{22} - c_{11}, \text{ and } H = c_{21}c_{12} - c_{11}c_{22}.$$

Since the parameters $(\alpha, h, \beta, \Lambda, \mu, r) \in \Omega_3$ and A^* varies in a small neighborhood of $A^* = 0$, and the roots of (6.3) are pair of complex conjugate numbers ω_1 and ω_2 denoted by

$$\omega_{1,2} = 1 + \frac{G(A_2 + A^*)}{2} \pm \frac{i(A_2 + A^*)}{2}\sqrt{4H - G^2},$$

we have

$$|\omega_{1,2}| = \sqrt{q(A^*)}, \qquad \frac{d|\omega_{1,2}|}{dA^*}\Big|_{A^*=0} = -\frac{G}{2} > 0.$$

In addition, it is required that, when $A^* = 0$, $\overline{\omega}^n$, $\omega^n \neq 1$ (n = 1, 2, 3, 4), which is equivalent to $p(0) \neq -2, 0, 1, 2$. Since $p^2(0) - 4q(0) < 0$ and q(0) = 1, we have $p^2(0) < 4$; then $p(0) \neq \pm 2$. It is only required that $p(0) \neq 0, 1$, which leads to

$$G^2 \neq 2H, 3H. \tag{6.4}$$

Therefore, the eigenvalues $\omega_{1,2}$ of fixed point (0, 0) of (6.2) does not lie on the intersection of the unit circle with the coordinate axes when $A^* = 0$.

Next, the normal form of model (6.2) when $A^* = 0$ is studied. Let $\theta = \text{Re}(\omega_{1,2})$, $\eta = \text{Im}(\omega_{1,2})$ and

$$T = \begin{pmatrix} c_{12} & 0\\ \theta - c_{11} & -\eta \end{pmatrix} = \begin{pmatrix} c_{11} & 0\\ \frac{-1}{2}K & \frac{-1}{2}K\sqrt{4K - K^2} \end{pmatrix}.$$
 (6.5)

Consider the translation below

$$\begin{pmatrix} X_n \\ Y_n \end{pmatrix} = T \begin{pmatrix} u_n \\ \upsilon_n \end{pmatrix}.$$
(6.6)

Taking T^{-1} on both sides of (6.2), we obtain

$$\begin{pmatrix} u_{n+1} \\ \upsilon_{n+1} \end{pmatrix} = \begin{pmatrix} \theta & -\eta \\ \eta & \theta \end{pmatrix} \begin{pmatrix} u_n \\ \upsilon_n \end{pmatrix} + \begin{pmatrix} f(u_n + \upsilon_n) \\ g(u_n + \upsilon_n) \end{pmatrix},$$
(6.7)

where

$$f(u_{n} + \upsilon_{n}) = c_{13} [(\theta - c_{11})u_{n}^{2} - \eta u_{n}\upsilon_{n}],$$

$$g(u_{n} + \upsilon_{n}) = [c_{13}(c_{11} - \theta) + c_{12}c_{23}] [\frac{(c_{11} - \theta)}{\eta}u_{n}^{2} + u_{n}\upsilon_{n}],$$

$$X_{n} = c_{12}u_{n}, \qquad Y_{n} = (\theta - c_{11})u_{n} - \eta\upsilon_{n}.$$

Now, we obtain

$$\frac{\partial^2 f}{\partial u_n^2} = 2c_{13}(\theta - c_{11}),$$

$$\frac{\partial^2 f}{\partial u_n \partial \upsilon_n} = -\eta c_{13},$$

$$\frac{\partial^2 f}{\partial \upsilon_n^2} = 0,$$

$$\frac{\partial^3 f}{\partial u_n^3} = \frac{\partial^3 f}{\partial u_n^2 \partial \upsilon_n} = \frac{\partial^3 f}{\partial u_n \partial \upsilon_n^2} = \frac{\partial^3 f}{\partial \upsilon_n^3} = 0,$$

$$\frac{\partial^2 g}{\partial u_n^2} = \frac{2(c_{11} - \theta)}{\eta} [c_{13}(c_{11} - \theta) + c_{12}c_{23}],$$

$$\frac{\partial^2 g}{\partial u_n \partial \upsilon_n} = c_{13}(c_{11} - \theta) + c_{12}c_{23},$$

$$\frac{\partial^2 g}{\partial \upsilon_n^2} = 0,$$

$$\frac{\partial^3 g}{\partial \upsilon_n^3} = \frac{\partial^3 g}{\partial u_n^2 \partial \upsilon_n} = \frac{\partial^3 g}{\partial u_n \partial \upsilon_n^2} = \frac{\partial^3 g}{\partial \upsilon_n^3} = 0.$$
(6.8)

According to the Neimark–Sacker bifurcation, the discriminatory quantity \hbar is given by

$$\hbar = -\operatorname{Re}\left[\frac{(1-2\omega)\overline{\omega}^{2}}{1-\omega}\xi_{11}\xi_{20}\right] - \frac{1}{2}|\xi_{11}|^{2} - |\xi_{02}|^{2} + \operatorname{Re}(\overline{\omega}\xi_{21}) \neq 0, \tag{6.9}$$

where

$$\begin{split} \omega &= \theta + i\eta, \\ \xi_{11} &= \frac{1}{4} \left[\frac{\partial^2 f}{\partial u_n^2} + \frac{\partial^2 f}{\partial v_n^2} + i \left(\frac{\partial^2 g}{\partial u_n^2} - \frac{\partial^2 g}{\partial v_n^2} \right) \right] \bigg|_{(u_n, v_n) = (0, 0)}, \end{split}$$

$$\begin{split} \xi_{20} &= \frac{1}{8} \left[\frac{\partial^2 f}{\partial u_n^2} - \frac{\partial^2 f}{\partial \upsilon_n^2} + 2 \frac{\partial^2 g}{\partial u_n \partial \upsilon_n} + i \left(\frac{\partial^2 g}{\partial u_n^2} - \frac{\partial^2 g}{\partial \upsilon_n^2} - 2 \frac{\partial^2 f}{\partial u_n \partial \upsilon_n} \right) \right] \Big|_{(u_n,\upsilon_n)=(0,0)}, \end{split}$$
(6.10)

$$\begin{split} \xi_{02} &= \frac{1}{8} \left[\frac{\partial^2 f}{\partial u_n^2} - \frac{\partial^2 f}{\partial \upsilon_n^2} - 2 \frac{\partial^2 g}{\partial u_n \partial \upsilon_n} + i \left(\frac{\partial^2 g}{\partial u_n^2} - \frac{\partial^2 g}{\partial \upsilon_n^2} + 2 \frac{\partial^2 f}{\partial u_n \partial \upsilon_n} \right) \right] \Big|_{(u_n,\upsilon_n)=(0,0)}, \end{aligned}$$

$$\begin{split} \xi_{21} &= \frac{1}{16} \left[\frac{\partial^3 f}{\partial u_n^3} + \frac{\partial^3 f}{\partial u_n \partial \upsilon_n^2} + \frac{\partial^3 g}{\partial u_n^2 \partial \upsilon_n} + \frac{\partial^3 g}{\partial \upsilon_n^3} \\ &\quad + i \left(\frac{\partial^3 g}{\partial u_n^3} + \frac{\partial^3 g}{\partial u_n \partial \upsilon_n^2} - \frac{\partial^3 f}{\partial u_n^2 \partial \upsilon_n} - \frac{\partial^3 f}{\partial \upsilon_n^3} \right) \right] \Big|_{(u_n,\upsilon_n)=(0,0)}. \end{split}$$

From the above analysis and the theorem in [47], Theorem 6.1 can be stated.

Theorem 6.1 If conditions (6.4) and (6.10) hold, then the system (3.7) undergoes Neimark– Sacker bifurcation at the positive fixed point $E_1(S^*, I^*)$ when the parameter A^* varies in the small neighborhood of A_2 . Furthermore if ($\hbar < 0$, $\hbar > 0$) then (an attracting, a repelling) invariant closed curve bifurcates from the fixed point $E_1(S^*, I^*)$ for ($A_2 > A^*$, $A_2 < A^*$), respectively.





7 Numerical examples

This section shows the bifurcation diagrams, phase portraits and maximum Lyapunov exponents for the model (3.7) to confirm the above theoretical analysis and to illustrate the complex dynamics of our model using numerical continuation. Bifurcation occurs when the stability of an equilibrium point changes [48]. In general, the dynamics of a discrete *SI* model with integer-order has been examined by Hu *et al.* [42]. As discussed earlier in Section 1, this paper focuses on varying the time step size parameter h and the fractional-order parameter α in the model (3.7), which can be seen as extension of the corresponding results given in [29, 42]. Based on the previous analysis, the parameters of the model (3.7) can be examined in the following two cases.





Case 1. Varying *h* in range $2 \le h \le 2.85$ and fixing $\Lambda = 3.5$, $\mu = 0.145$, r = 0.12, $\beta = 0.09$, $\alpha = 0.99$ with initial conditions (S_0 , I_0) = (3.1, 7.8).

Case 2. Varying α in range $0.8 \le \alpha \le 0.99$ and fixing $\Lambda = 1.5$, $\mu = 0.2$, r = 0.3, $\beta = 0.1$, h = 8.1, $(S_0, I_0) = (5.1, 0.9)$.

In Case 1, the basic reproduction number $\Re_0 = 8.198 > 1$, thus the model (3.7) has one positive endemic equilibrium point $E_1(2.944, 11.596)$. Since $\Delta = 0.307 > 0$ and $h_1 = 2.305$,



according to Theorem 4.4, E_1 is asymptotically stable (sink) when h < 2.305. When $h = h_1$, flip bifurcation emerges from the equilibrium point E_1 with $\chi_1 = -0.871$, $\chi_2 = 0.015 > 0$ and $(\alpha, h, \beta, \Lambda, \mu, r) = (0.99, 2.305, 0.09, 3.5, 0.145, 0.12) \in \Omega_1$. The occurrence of these bifurcations is illustrated in Figure 1(a)–(c). These figures show that E_1 is stable when h < 2.305 and loses its stability through the flip bifurcation when h = 2.305. Period-2, 4, 8, 16 orbits appear as h increases in the range when $h \in (2.305, 2.83)$. The phase portraits for $h \in (2.305, 2.83)$ are plotted in Figure 2(a)–(d) to illustrate these observations further. The emergence of period-2, 4, 8 and 16 orbits are observed when h = 2.5, 2.75, 2.8 and 2.82 in Figure 2(a)–(d). Some interesting phenomena are also seen when h increases further: for instance, when h = 2.9 (Figure 2(e)), period-12 orbit appears. The occurrence of chaotic regions is also observed in Figure 1(a)–(c): these phenomena can be illustrated by the phase portrait in Figure 2(f) (*e.g.* when h = 2.95). Maximal Lyapunov exponents (LEs) are





computed in Figure 3 corresponding to observations in Figure 1(a)–(c). It is observed that some LE values are positive and some are negative, so there exist stable fixed points or stable period windows in the chaotic regions. Generally, a positive LE is considered to be one of the characteristics which imply the existence of chaos (*e.g.* when h > 2.83).

In Case 2, the basic reproduction number $\Re_0 = 1.5 > 1$, thus, model (3.7) has one positive endemic equilibrium point $E_1(5,1)$. Since $\Delta = -0.11 < 0$, according to Theorem 4.4 E_1 is asymptotically stable (sink) when $\alpha < 0.826$ and unstable when $\alpha > 0.826$. The Neimark–Sacker (N-S) bifurcation emerges from E_1 at $\alpha = 0.826$ with $\hbar = -0.778 < 0$ and $(\alpha, h, \beta, \Lambda, \mu, r) = (0.826, 8.1, 0.1, 1.5, 0.2, 0.3) \in \Omega_3$. The occurrence of an N-S bifurcation is illustrated in Figure 4(a)–(c). Figure 4 shows that E_1 is stable for $\alpha < 0.826$ and loses its stability through an N-S bifurcation at α = 0.826. Attracting invariant circle appears as fractional-order parameter increases in the range of $\alpha \in (0.826, 0.87)$. The phase portraits for various α -values corresponding to Figure 4 are plotted in Figure 5(a)–(i) to illustrate these observations. Furthermore, the quasi-periodic orbits ($\alpha = 0.913$) and periodic-13 orbits ($\alpha = 0.921$) are observed within the chaotic regions in Figure 5(f) and (g), respectively. Attracting chaotic sets are also seen when α increases further and these observations are plotted in Figure 5(h)-(i). Maximal Lyapunov exponents corresponding to observations in Figure 4(a)-(c) are computed in Figure 6. It is observed that some LE values are positive and some are negative. So there exist stable fixed points or stable period windows in the chaotic regions (*e.g.* when $\alpha > 0.921$).

8 Discussion and conclusion

A new discrete-time SI epidemic model has been discussed in this paper. Such a discretetime model is obtained from the discretization of the fractional-time SI model. The discretization process provides crucial terms such as h (time step parameter) and α (fractional-order parameter), which are then varied in order to describe the dynamical behaviors of the model. As h and α are varied, the model exhibits several complicated dynamical behaviors including the emergence of flip and N-S bifurcations, period-2, 4, 8, 12, 13, 16 orbits, quasi-periodic orbits, attracting invariant circle and chaotic sets. Analytically, necessary and sufficient conditions on the parameters for the occurrence of flip and N-S bifurcations are derived. Moreover, numerical continuation is carried out to illustrate the validity of the analytical results and it is observed that both numerical and analytical findings are in good agreement. In conclusion, the proposed fractional-time SI model can engender more complex dynamical behaviors than its continuous model counterpart. Studying the dynamical behaviors of the full avian-human influenza SIR epidemic model described by a discrete-time with fractional-order version will be our future work.

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Competing interests

The authors declare that they have no competing interests.

Authors' contributions

The main idea of this paper was proposed by MAMA. The manuscript prepared initially and all steps of the proof performed by MAMA, All, FAA, and MHM. All authors read and approved the final manuscript.

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