

Mathematics

Chaotic dynamics of the sugarcane borer-two parasitoid agroecosystem with seasonality

Dinâmica caótica do agroecossistema da broca da cana de açúcar com dois parasitóides e sazonalidade

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ABSTRACT

Sugarcane production is a significant and profitable agribusiness sector in many countries. Nevertheless, this industry suffers significant losses from sugarcane pests, among which the most important is the sugarcane borer (*Diatraea saccharalis*). This pest population is hard to control due to its different life stages, thus biological control (with more than one predator species) can be applied. Therefore, in this work, we present and analyze a mathematical model that describes the dynamics of the sugarcane borer and its two parasitoids at the egg stage (*Trichogramma galloi*) and larval stage (*Cotesia flavipes*). First, a host-parasitoid model is used to obtain the population dynamics, which also considers the influence of seasonal variations. Then, system simulations and bifurcation diagrams show that the introduction of seasonality perturbations causes complex dynamics and results in limit cycles and strange attractors.

Keywords: Sugarcane borer; Parasitoids; Dynamics; Seasonality; Chaos

RESUMO

A produção de cana-de-açúcar é um setor significativo e lucrativo do agronegócio em muitos países. No entanto, essa indústria sofre perdas significativas com as pragas da cana-de-açúcar, dentre as quais a mais importante é a broca da cana-de-açúcar (*Diatraea saccharalis*). Essa população de pragas é difícil de ser controlada devido às suas diferentes fases de vida, portanto o controle biológico (com mais de uma espécie de predador) pode ser aplicado. Neste trabalho, apresentamos e analisamos um modelo matemático que descreve a dinâmica da broca da cana-de-açúcar e seus dois parasitoides, nos estágios de ovos (*Trichogramma galloi*) e de larvas (*Cotesia flavipes*). Primeiro, um modelo hospedeiro-parasitoide é usado para obter a dinâmica populacional, que também considera a influência das variações sazonais. Em

seguida, simulações do sistema e diagramas de bifurcação mostram que a introdução de perturbações de sazonalidade causam dinâmicas complexas e resultam em ciclos limite e atratores estranhos.

Palavras-chave: Broca da cana; Parasitoides; Dinâmica; Sazonalidade; Caos

1 INTRODUCTION

Sugarcane is produced in the tropics, as a global commodity, it is used in the manufacture of sweeteners, biofuels and a growing range of bioproducts (including bioplastics) (Commodities Jurisdictions (2022)). Several pests affect the sugarcane's production, specially the sugarcane borer *D. sacharalis* (Parra et al. (2002); Parra and Coelho (2019)). Significant damages of the plant are caused by this peculiar insect, that lays eggs on the surface of the sugarcane leaves, while the larvae lives inside the sugarcane stalk, carving internal galleries. Because the larvae live hidden from the surface, the pesticide control becomes inefficient and biological control is the alternative for this case.

The sugar borer infestation can cause direct and indirect damages. Direct damage includes the breaking of stalks and the demise of young plants, often caused by the condition referred to as "dead heart", resulting from the destruction of the apical meristem. Additionally, indirect damage occurs as the borer creates openings for fungi to enter, triggering the red rot disease in the stalk. This disease not only compromises the quality of production but also reduces sugar content in plants. Quantities of sugar losses (per ton of processed sugarcane) caused by the intensities of *D. saccharalis* infestations can be seen in the technical report Nava et al. (2009). In this technical report, the authors also indicate that the level of economic damage to the sugarcane borer is 1000 larvae/ hectare, which in other studies such as Rafikov and Silveira (2013) and Rafikov and Silveira (2014), can reach 2500 larvae/ hectare.

The damage caused by the sugarcane borer to sugarcane crops has been analyzed and debated by several researchers. In the search to minimize losses in the production of alcohol and sugar, the researchers' focus is on combating this pest. In the works of Belezini Vinha et al. (2019); Borges Filho et al. (2018,1); Parra et al. (2021), one can notice a concern with understanding the behavior of the sugarcane borer and proposing biological control strategies.

In biological control, the pest populations are reduced by the insertion of their natural enemies(predators, parasitoids and pathogens) in the environment. The sugarcane borer larvae population has been controlled since the 1970s by the parasitoid, *C. flavipes*. Recently, the *T. galloi* parasitoid is a new alternative for biological control of its egg population Parra et al. (2002); Parra and Coelho (2019). The works Gámez et al. (2009,1); Venturino (2007); Venturino et al. (2008) presented mathematical modeling in biological control applications of prey-predator and host-parasitoid.

In biological systems with more than two species, prey-predator interactions become complex and consequently harder to model prey invasions. Mathematical modeling can be an important feature providing information about the natural systems' stability (Goh (2012)), along with computational simulations, revealing the behavior of these complex systems and understanding how the prey interacts with other species in the environment.

Mathematical models for biological control, with only one predator population, are considered in Rafikov and Holanda Limeira (2012) and Rafikov and Silveira (2014) for egg and larval parasitoid, respectively. The mathematical model of interactions between the sugarcane borer and its eggs and larval parasitoids was proposed by Rafikov and Silveira (2013). In Molnár et al. (2016), this model was used for the formulation of some scenarios of biological pest control. In both publications, the populations are described by four population compartments: sugarcane borer's eggs population, sugarcane borer's larvae population, *T. galloi*(eggs parasitoid) population, *C. Flavipes*(larval parasitoid) population.

As pointed out by Rinaldi et al. (1993), periodic external forces are of great importance in ecological systems since environments of the population communities vary periodically. There are many systems that have a very simple dynamic behavior in the constant parameter case but become very complex (multiplicity of attractors, catastrophes, and chaos) when they are periodically perturbed.

Seasonality in biological systems has already been addressed in several works, such as Altizer et al. (2006); Bezerra et al. (2021); Gakkhar et al. (2003); Gakkhar and Naji (2003); Rinaldi et al. (1993); Stollenwerk et al. (2017); White and Hastings (2020); Zhang et al. (2012). Meanwhile, the inclusion of two parasitoids (eggs and larval) populations, as well as seasonal variations in the sugarcane borer agroecosystem

dynamics are novelties. In Bezerra et al. (2021), models sugarcane borer and its larval parasitoid (*C. flavipes*) interaction, considering the influence of the seasonal variations on the dynamics of the system. Their results show that this variation generates chaotic dynamics in the system.

The system model in the present paper is an extension of the mathematical model in Rafikov and Silveira (2013), with the addition of the parasitized egg and larvae population of the sugarcane borer. This addition improves the estimation and observation of the system parameters, as it is much easier to monitor sugarcane borer's parasitized eggs and parasitized larvae than adult parasitoid population in real conditions. Moreover, seasonality variations are introduced into the system dynamics resulting in chaotic behavior.

The paper is organized as follows. In Section 2, our six-dimensional, continuous-time, dynamical system is proposed. Section 3 is dedicated to the study of the system's local and global dynamics. Seasonal dynamics of the sugarcane borer parasitoid agroecosystem is considered in Section 4. Section 5 discusses the results of previous Sections and concludes this paper.

2 MATHEMATICAL MODEL

The proposed continuous-time mathematical model describes the interactions between the sugarcane borer, its egg and larval parasitoid, considering six population densities: the un-parasitized egg population density of the sugarcane borer, x_1 ; the parasitized egg population density of the sugarcane borer, x_2 ; the density of the adult egg parasitoid *T. galloi*, x_3 ; the unparasitized larvae density of the sugarcane borer, x_4 ; the parasitized larvae density of the sugarcane borer, x_5 ; the density of the adult larval parasitoid *C. flavipes*, x_6 . So, the mathematical model has the following form:

$$\begin{aligned}
\frac{dx_1}{dt} &= rx_1 \left(1 - \frac{x_1}{K}\right) - m_1x_1 - n_1x_1 - \alpha x_1x_3, \\
\frac{dx_2}{dt} &= \alpha x_1x_3 - m_2x_2 - n_2x_2, \\
\frac{dx_3}{dt} &= \gamma_1 n_2x_2 - m_3x_3, \\
\frac{dx_4}{dt} &= n_1x_1 - m_4x_4 - n_3x_4 - \beta x_4x_6, \\
\frac{dx_5}{dt} &= \beta x_4x_6 - m_5x_5 - n_4x_5, \\
\frac{dx_6}{dt} &= \gamma_2 n_4x_5 - m_6x_6.
\end{aligned} \tag{1}$$

In the system of differential equations (1), the 16 parameters are defined as follows. r is the intrinsic oviposition rate of female sugarcane borer; K is the potential maximum of oviposition rate of female sugarcane borer; m_1, m_2, m_3, m_4, m_5 and m_6 are the mortality rates of the un-parasitized egg, parasitized egg, egg parasitoid, un-parasitized larvae, parasitized larvae and larvae parasitoid populations, respectively; n_1 is the fraction of the sugarcane borer larvae population which emerges from the eggs per unit of time; n_2 is the fraction of the parasitized egg population from which larval parasitoids emerge in a time unit; n_3 is the fraction of the un-parasitized sugarcane borer larvae from which pupae emerge in a time unit; n_4 is the fraction of the parasitized sugarcane borer larvae from which larvae parasitoids emerge in a time unit; α and β are the intrinsic parasitism rate of the egg and larvae parasitoids, respectively; γ_1 and γ_2 are numbers of adult parasitoids which emerge from a unit of parasitized eggs and larvae, respectively.

3 SYSTEM EQUILIBRIUM STATES AND THEIR STABILITY

3.1 Equilibrium states

To obtain the equilibrium points, we equal the right-hand sides of the system (1) to zero. We obtain five following equilibrium by sequences of $x_i, i = 1, 2, \dots, 6$.

- Extinction of all populations: $E_1 = (0, 0, 0, 0, 0, 0)$.
- Extinction of all parasitoid and parasitized populations:
 $E_2 = \left(\frac{K}{r}(r - m_1 - n_1), 0, 0, \frac{Kn_1(r - m_1 - n_1)}{r(m_4 + n_3)}, 0, 0\right)$.
- Extinction of the egg parasitoid and parasitized egg populations:

$$E_3 = \left(\frac{K}{r}(r - m_1 - n_1), 0, 0, p_4^*, p_5^*, p_6^* \right).$$

- Extinction of the larvae parasitoid and parasitized larvae populations:

$$E_4 = (x_1^*, x_2^*, x_3^*, q^*, 0, 0).$$

- Coexistence of all populations: $E_5 = (x_1^*, x_2^*, x_3^*, x_4^*, x_5^*, x_6^*)$.

The values cited above are given as follows.

$$\begin{aligned} x_1^* &= \frac{m_3(m_2 + n_2)}{\alpha\gamma_1 n_2}, & x_3^* &= \frac{1}{\alpha} \left[r \left(1 - \frac{m_3(m_2 + n_2)}{\alpha\gamma_1 n_2 K} \right) - m_1 - n_1 \right], & x_2^* &= \frac{m_3}{\gamma_1 n_2} x_3^*, \\ x_4^* &= \frac{m_6(m_5 + n_4)}{\beta\gamma_2 n_4}, & x_5^* &= \frac{m_3 n_1 (m_2 + n_2)}{\alpha\gamma_1 n_2 (m_5 + n_4)} - \frac{m_6(m_4 + n_3)}{\beta\gamma_2 n_4}, & x_6^* &= \frac{\gamma_2 n_4}{m_6} x_5^*, \\ p_4^* &= \frac{m_6(m_5 + n_4)}{\beta\gamma_2 n_4}, & p_5^* &= \frac{n_1}{m_5 + n_4} \left[\frac{K}{r}(r - m_1 - n_1) \right] - \frac{m_6(m_4 + n_3)}{\beta\gamma_2 n_4}, & p_6^* &= \frac{\gamma_2 n_4}{m_6} p_5^*, \\ q^* &= \frac{n_1}{m_4 + n_3} x_1^*. \end{aligned}$$

Since we are modelling a biological system, where the dependent variables are populations, the space phase includes only positive or zero values for all coordinates, thus defining their biological viability. When analyzing, in next section, each one of these equilibrium points, conditions will be set for their biological viability.

3.2 Local stability analysis of equilibrium points

In order to study the local the stability of these equilibrium states, system (1) is linearized in a small neighborhood of each equilibrium state, and the Jacobian matrix being computed as:

$$J = \begin{bmatrix} a_{11} & 0 & a_{13} & 0 & 0 & 0 \\ a_{21} & a_{22} & a_{23} & 0 & 0 & 0 \\ 0 & a_{32} & a_{33} & 0 & 0 & 0 \\ a_{41} & 0 & 0 & a_{44} & 0 & a_{46} \\ 0 & 0 & 0 & a_{54} & a_{55} & a_{56} \\ 0 & 0 & 0 & 0 & a_{65} & a_{66} \end{bmatrix}, \quad (2)$$

in which:

$$a_{11} = r - \frac{2rx_1}{K} - m_1 - n_1 - \alpha x_3, \quad a_{13} = -\alpha x_1, \quad a_{21} = \alpha x_3, \quad a_{22} = -m_2 - n_2,$$

$$a_{23} = \alpha x_1, \quad a_{32} = \gamma_1 n_2, \quad a_{33} = -m_3, \quad a_{41} = n_1, \quad a_{44} = -m_4 - n_3 - \beta x_6,$$

$$a_{46} = -\beta x_4, \quad a_{54} = \beta x_6, \quad a_{55} = -m_5 - n_4, \quad a_{56} = \beta x_4, \quad a_{65} = \gamma_2 n_4, \quad a_{66} = -m_6.$$

The matrix (2) can be written as a block matrix:

$$\begin{bmatrix} A & 0 \\ C & B \end{bmatrix}, \tag{3}$$

in which:

$$A = \begin{bmatrix} a_{11} & 0 & a_{13} \\ a_{21} & a_{22} & a_{23} \\ 0 & a_{32} & a_{33} \end{bmatrix}, \quad C = \begin{bmatrix} a_{41} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} a_{44} & 0 & a_{46} \\ a_{54} & a_{55} & a_{56} \\ 0 & a_{65} & a_{66} \end{bmatrix},$$

and 0 is a matrix with all elements equal to zero.

Matrices of the form (3) are called block-lower-triangular. The full explanation on the block-triangular matrix determinant computation method can be found in Gantmacher (1960), and briefly described below:

If D is a block-triangular matrix, then the determinant of the matrix is equal to the product of determinant of diagonal cells:

$$\det(D) = \det(D_{11}) \det(D_{22}) \cdots \det(D_{nn}). \tag{4}$$

The matrix $D = J - \lambda I$, where I is the identity matrix of dimensions $n \times n$, is block-lower-triangular too. Using the above mentioned rule, the characteristic equation is given as follows:

$$\det(D) = \det(A - \lambda I) \det(B - \lambda I) = 0. \tag{5}$$

Hence, the characteristic equation (5) is given by:

$$\begin{vmatrix} a_{11} - \lambda & 0 & a_{13} \\ a_{21} & a_{22} - \lambda & a_{23} \\ 0 & a_{32} & a_{33} - \lambda \end{vmatrix} \begin{vmatrix} a_{44} - \lambda & 0 & a_{46} \\ a_{54} & a_{55} - \lambda & a_{56} \\ 0 & a_{65} & a_{66} - \lambda \end{vmatrix} = 0. \quad (6)$$

From (6), we get that:

$$\begin{vmatrix} a_{11} - \lambda & 0 & a_{13} \\ a_{21} & a_{22} - \lambda & a_{23} \\ 0 & a_{32} & a_{33} - \lambda \end{vmatrix} = 0, \quad \text{or} \quad (7)$$

$$\begin{vmatrix} a_{44} - \lambda & 0 & a_{46} \\ a_{54} & a_{55} - \lambda & a_{56} \\ 0 & a_{65} & a_{66} - \lambda \end{vmatrix} = 0. \quad (8)$$

Now, applying this rule to the equilibrium points found above, we obtain for the equilibrium point $E_1 = (0, 0, 0, 0, 0, 0)$, the matrix (2) has a triangular form, and the eigenvalues are given by:

$\lambda_1 = r - m_1 - n_1$, $\lambda_2 = -m_2 - n_2$, $\lambda_3 = -m_3$, $\lambda_4 = -m_4 - n_3$, $\lambda_5 = -m_5 - n_4$, and $\lambda_6 = -m_6$. Therefore, it follows that equilibrium E_1 is asymptotically stable if $r < m_1 + n_1$.

For the equilibrium $E_2 = \left(\frac{K}{r}(r - m_1 - n_1), 0, 0, \frac{Kn_1(r - m_1 - n_1)}{r(m_4 + n_3)}, 0, 0 \right)$, which is biologically viable if $r > m_1 + n_1$, the parameters of the characteristic equation (5) are given as:

$$\begin{aligned} a_{11} &= -(r - m_1 - n_1), \quad a_{13} = -\frac{\alpha K}{r}(r - m_1 - n_1), \quad a_{21} = 0, \quad a_{22} = -m_2 - n_2, \\ a_{23} &= \frac{\alpha K}{r}(r - m_1 - n_1), \quad a_{32} = \gamma_1 n_2, \quad a_{33} = -m_3, \quad a_{44} = -m_4 - n_3, \\ a_{46} &= -\frac{\beta K n_1 (r - m_1 - n_1)}{r(m_4 + n_3)}, \quad a_{54} = 0, \quad a_{55} = -m_5 - n_4, \quad a_{56} = \frac{\beta K n_1 (r - m_1 - n_1)}{r(m_4 + n_3)}, \\ a_{65} &= \gamma_2 n_4, \quad a_{66} = -m_6. \end{aligned}$$

From (7) and (8), we obtain:

$$(a_{11} - \lambda)[\lambda^2 - (a_{22} + a_{33})\lambda + a_{22}a_{33} - a_{23}a_{32}] = 0, \quad \text{or} \quad (9)$$

$$(a_{44} - \lambda)[\lambda^2 - (a_{55} + a_{66})\lambda + a_{55}a_{66} - a_{56}a_{65}] = 0. \quad (10)$$

The Routh-Hurwitz criterion states that the eigenvalues of the second-degree polynomial have negative real parts if, and only if, both coefficients are positive. Analyzing (7) and (8), we can conclude that when

$$(a) a_{11} < 0, \quad (b) a_{22} + a_{33} - a_{23}a_{32} > 0, \quad (c) a_{55} + a_{66} - a_{56}a_{65} > 0,$$

then all the eigenvalues of the equation (6) have negative real parts.

From conditions (a), (b) and (c), we obtain:

$$\alpha < \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 K(r - m_1 - n_1)}, \quad (11)$$

$$\beta < \frac{rm_6(m_4 + n_3)(m_5 + n_4)}{\gamma_2 n_1 n_4 K(r - m_1 - n_1)}, \quad (12)$$

$$r > m_1 + n_1. \quad (13)$$

Similarly, the equilibrium point E_2 is asymptotically stable if, and only if, the inequalities (11), (12) and (13) are satisfied.

The equilibrium point $E_3 = \left(\frac{K}{r}(r - m_1 - n_1), 0, 0, p_4^*, p_5^*, p_6^*\right)$ is biologically viable if:

$$\beta > \frac{rm_6(m_4 + n_3)(m_5 + n_4)}{\gamma_2 n_1 n_4 K(r - m_1 - n_1)}, \quad (14)$$

$$r > m_1 + n_1.$$

The characteristic equation at E_3 can be written in the form (6), where:

$$a_{11} = -(r - m_1 - n_1), \quad a_{13} = -\frac{\alpha K}{r}(r - m_1 - n_1), \quad a_{21} = 0, \quad a_{22} = -m_2 - n_2,$$

$$a_{23} = \frac{\alpha K}{r}(r - m_1 - n_1), \quad a_{32} = \gamma_1 n_2, \quad a_{33} = -m_3, \quad a_{44} = -m_4 - n_3 - \beta p_6^*,$$

$$a_{46} = -\beta p_4^*, \quad a_{54} = \beta p_6^*, \quad a_{55} = -m_5 - n_4, \quad a_{56} = \beta p_4^*, \quad a_{65} = \gamma_2 n_4, \quad a_{66} = -m_6.$$

The parameter values of determinant (7) are the same as E_2 , and the conditions (11) and (13) are satisfied.

Considering the determinant (8)

$$\begin{vmatrix} a_{44} - \lambda & 0 & a_{46} \\ a_{54} & a_{55} - \lambda & a_{56} \\ 0 & a_{65} & a_{66} - \lambda \end{vmatrix} = 0, \quad (15)$$

we have

$$\lambda^3 + b_1\lambda^2 + b_2\lambda + b_3 = 0, \quad (16)$$

where

$$\begin{aligned} b_1 &= m_4 + n_3 + \beta p_6^* > 0, & b_2 &= (m_4 + n_3 + \beta p_6^*)(m_5 + n_4) > 0, \\ b_3 &= \beta m_6(m_5 + n_4)p_6^* > 0, & b_1b_2 - b_3 &> 0. \end{aligned} \quad (17)$$

Therefore, we obtain that the equilibrium point E_3 is asymptotically stable if, and only if, the inequalities (11), (14) and (13) are satisfied.

The equilibrium point $E_4 = (x_1^*, x_2^*, x_3^*, q^*, 0, 0)$ is biologically viable if

$$\alpha > \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 K(r - m_1 - n_1)}, \quad (18)$$

$$r > m_1 + n_1.$$

The characteristic equation at E_4 can be written in the form (6) where:

$$\begin{aligned} a_{11} &= -\frac{rx_1^*}{K}, & a_{13} &= -ax_1^*, & a_{21} &= ax_3^*, & a_{22} &= -m_2 - n_2, \\ a_{23} &= ax_1^*, & a_{32} &= \gamma_1 n_2, & a_{33} &= -m_3, & a_{44} &= -m_4 - n_3, \\ a_{46} &= -\beta q^*, & a_{54} &= 0, & a_{55} &= -m_5 - n_4, & a_{56} &= \beta q^*, & a_{65} &= \gamma_2 n_4, & a_{66} &= -m_6. \end{aligned} \quad (19)$$

Considering the determinant (7) with parameter values (19) we have:

$$\lambda^3 + c_1\lambda^2 + c_2\lambda + c_3 = 0. \quad (20)$$

where

$$c_1 = m_2 + m_3 + n_2 + \frac{rx_1^*}{K} > 0, \quad c_2 = (m_2 + m_3 + n_2)\frac{rx_1^*}{K} > 0, \quad c_3 = \alpha m_3(m_2 + n_2)x_3^* > 0.$$

From $c_1c_2 - c_3 > 0$, we obtain:

$$\alpha < \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 z K}, \tag{21}$$

where

$$z = -\frac{h_1}{2} + \sqrt{\frac{h_1^2}{4} + h_2},$$

$$h_1 = m_2 + m_3 + n_2 + \frac{m_3(m_2 + n_2)}{m_2 + m_3 + n_2},$$

$$h_2 = \frac{m_3(m_2 + n_2)(r - m_1 - n_1)}{m_2 + m_3 + n_2}.$$

Considering the determinant (8) with parameter values (19) we have:

$$(a_{44} - \lambda)[\lambda^2 - (a_{55} + a_{66})\lambda + a_{55}a_{66} - a_{56}a_{65}] = 0, \tag{22}$$

where $a_{44} = -m_4 - n_3 < 0$, $-(a_{55} + a_{66}) > 0$.

From $a_{55}a_{66} - a_{56}a_{65} > 0$ we obtain:

$$\beta < \frac{m_6(m_5 + n_4)}{\gamma_2 n_4 q^*} = \frac{\alpha \gamma_1 n_2 m_6 (m_4 + n_3)(m_5 + n_4)}{\gamma_2 n_1 n_4 m_3 (m_2 + n_2)}. \tag{23}$$

The equilibrium point E_4 is asymptotically stable if, and only if, the following inequalities are satisfied:

$$\frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 K (r - n_1 - m_1)} < \alpha < \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 z K}, \tag{24}$$

$$\beta < \frac{\alpha \gamma_1 n_2 m_6 (m_4 + n_3)(m_5 + n_4)}{\gamma_2 n_1 n_4 m_3 (m_2 + n_2)}. \tag{25}$$

Consider the equilibrium point $E_5 = (x_1^*, x_2^*, x_3^*, x_4^*, x_5^*, x_6^*)$ where

$$\begin{aligned} x_1^* &= \frac{m_3(m_2 + n_2)}{\alpha\gamma_1 n_2}, x_3^* = \frac{1}{\alpha} \left[r \left(1 - \frac{m_3(m_2 + n_2)}{\alpha\gamma_1 n_2 K} - m_1 - n_1 \right) \right], x_2^* = \frac{m_3}{\gamma_1 x_2} x_3^*, \\ x_4^* &= \frac{m_6(m_5 + n_4)}{\beta\gamma_2 n_4}, x_5^* = \frac{m_3 n_1 (m_2 + n_2)}{\alpha\gamma_1 n_2 (m_5 + n_4)} - \frac{m_6(m_4 + n_3)}{\beta\gamma_2 n_4}, x_6^* = \frac{\gamma_2 n_4}{m_6} x_5^*. \end{aligned} \quad (26)$$

The equilibrium point E_5 is biologically viable if $x_3^* > 0$ and $x_5^* > 0$. So, we get:

$$\begin{aligned} \alpha &> \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 K (r - n_1 - m_1)}, \\ \beta &> \frac{\alpha\gamma_1 n_2 m_6 (m_4 + n_3) (m_5 + n_4)}{\gamma_2 n_1 n_4 m_3 (m_2 + n_2)}, \\ r &> m_1 + n_1. \end{aligned} \quad (27)$$

The characteristic equation at E_5 can be written in the form (6) where:

$$\begin{aligned} a_{11} &= -\frac{rx_1^*}{K}, \quad a_{13} = -ax_1^*, \quad a_{21} = ax_3^*, \quad a_{22} = -m_2 - n_2, \\ a_{23} &= ax_1^*, \quad a_{32} = \gamma_1 n_2, \quad a_{33} = -m_3, \quad a_{44} = -m_4 - n_3 - \beta x_6^*, \\ a_{46} &= -\beta x_4^*, \quad a_{54} = \beta x_6^*, \quad a_{55} = -m_5 - n_4, \quad a_{56} = \beta x_4^*, \quad a_{65} = \gamma_2 n_4, \quad a_{66} = -m_6. \end{aligned} \quad (28)$$

The parameter values of determinant (7) are the same of E_4 , and the inequality (21) is satisfied.

Considering the determinant (8), we obtain:

$$\lambda^3 + g_1 \lambda^2 + g_2 \lambda + g_3 = 0, \quad (29)$$

where

$$\begin{aligned} g_1 &= m_4 + m_5 + m_6 + n_3 + n_4 + \beta x_6^* > 0, \\ g_2 &= (m_4 + n_3 + \beta x_6^*)(m_5 + n_4 + m_6) > 0, \\ g_3 &= \beta m_6 (m_5 + n_4) x_6^* > 0, \quad g_1 g_2 - g_3 > 0. \end{aligned} \quad (30)$$

Therefore, we obtain that equilibrium point E_5 is asymptotically stable if, and only if, the following inequalities are satisfied:

$$\frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 K(r - m_1 - n_1)} < \alpha < \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 z K}, \tag{31}$$

$$\beta > \frac{\alpha \gamma_1 n_2 m_6 (m_4 + n_3) (m_5 + n_4)}{\gamma_2 n_1 n_4 m_3 (m_2 + n_2)}. \tag{32}$$

We can now summarize the results of this local stability analysis, after defining the following dimensionless parameters related to our resulting conditions (regions in the parameter space) for biological viability (b. v.) and for local stability (l.s.) of each equilibrium point:

$$A_1 \equiv \frac{r}{m_1 + n_1}; A_2 \equiv \frac{\alpha \gamma_1 n_2 K(r - m_1 - n_1)}{rm_3(m_2 + n_2)}; A_3 \equiv \frac{\beta \gamma_2 n_1 n_4 K(r - m_1 - n_1)}{rm_6(m_4 + n_3)(m_5 + n_4)}; \tag{33}$$

$$A_4 \equiv \frac{\beta \gamma_2 n_1 n_4 m_3 (m_2 + n_2)}{\alpha \gamma_1 n_2 m_6 (m_4 + n_3) (m_5 + n_4)} = \frac{A_3}{A_2}; A_5 \equiv \frac{\alpha \gamma_1 n_2 z K}{rm_3(m_2 + n_2)} = \frac{A_2 z}{r - m_1 - n_1}.$$

With these dimensionless parameters, whose critical value 1 is associated with a bifurcation in the behavior of the system, all conditions previously deduced can be presented in a summary and complete form as shown in Table 1.

Table 1 – Local stability and biological viability of the system equilibria, in which b.v. means biologically viable, l.s. means locally stable and un. means unstable

		E_1		E_2		E_3		E_4		E_5			
$A_1 < 1$		b.v.	l.s.	Not b.v.		Not b.v.		Not b.v.		Not b.v.			
$A_1 > 1$	$A_2 < 1$	$A_3 < 1$		b.v.	un.	b.v.	l.s.	Not b.v.		Not b.v.			
		$A_3 > 1$		b.v.	un.	b.v.	un.	b.v.	l.s.	Not b.v.			
	$A_2 > 1$	$A_5 < 1$	$A_4 < 1$	b.v.	un.	b.v.	un.	b.v.	un.	b.v.	l.s.	Not b.v.	
			$A_4 > 1$	b.v.	un.	b.v.	un.	b.v.	un.	b.v.	un.	b.v.	l.s.
	$A_5 > 1$	$A_4 > 1$	b.v.	un.	b.v.	un.	b.v.	un.	b.v.	un.	b.v.	un.	

3.3 Global stability analysis of the coexistence equilibrium

From local stability analysis, we get the equilibrium point E_5 is the only one with no null population density. Therefore this is the point of interest for the forthcoming global stability analysis.

We define a Lyapunov function as follows:

$$V(x_1, x_2, x_3, x_4, x_5, x_6) = \int_{x_1^*}^{x_1} \frac{y - x_1^*}{y} dy + \sum_{i=2}^6 \frac{(x_i - x_1^*)^2}{2}. \quad (34)$$

At the equilibrium point E_5 , V is zero, and it is positive for all other biologically viable equilibria. The function is also radially unbounded, i.e., $V \rightarrow \infty$ when $x \rightarrow \infty$.

We can write the time derivative of V along (1) as

$$\dot{V} = e^T P e, \quad (35)$$

where the matrix P is

$$\begin{bmatrix} -\frac{r}{K} & 0 & -\alpha & 0 & 0 & 0 \\ \alpha x_3 & -m_2 - n_2 & \alpha x_1^* & 0 & 0 & 0 \\ 0 & \gamma_1 n_2 & -m_3 & 0 & 0 & 0 \\ n_1 & 0 & 0 & -m_4 - n_3 - \beta x_6 & 0 & -\beta x_4^* \\ 0 & 0 & 0 & \beta x_6 & -m_5 - n_4 & \beta x_4^* \\ 0 & 0 & 0 & 0 & \gamma_2 n_4 & -m_6 \end{bmatrix},$$

and the elements of vector e are: $e_i = x_i - x_i^*$, $i = 1, 2, \dots, 6$.

If conditions (31) and (32) are satisfied then the matrix P in (35) is negative definite, as is the time derivative of V along the trajectories of (1), and consequently the equilibrium point E_5 is globally asymptotically stable.

3.4 Hopf bifurcation analysis

From Table 1, it is evident the role of the dimensionless parameter A_5 in determining the region in the parameter space in which each one of the equilibrium points E_4 and E_5 is asymptotically stable. Furthermore, from the definition of A_5 in (33), it can be seen that the condition $A_5 < 1$ can be written equivalently in the form of $\alpha < \alpha_c$ where the critical value α_c is defined by

$$\alpha_c \equiv \frac{r m_3 (m_2 + n_2)}{\gamma_1 n_2 z K} \quad (36)$$

In (36) we observe that the condition is exactly the same one as we got in (21), from the characteristic (20), for the local stability of the equilibria E_4 and E_5 .

When $A_5 > 1$, that is, $\alpha > \alpha_c$, the positive coexistence equilibrium E_5 becomes unstable and a Hopf bifurcation occurs.

Now we can analyze the bifurcation of the model (1) assuming α as the bifurcation parameter and considering only three first equations of the system (1) which are not dependent on the variables x_4, x_5 and x_6 . The traditional Hopf bifurcation criterion is stated in terms of the properties of the eigenvalues. Alternatively, Liu (1994) presented a criterion of Hopf bifurcation without using the eigenvalues of the characteristic equation. Liu's approach is the one that is applied in the present Hopf bifurcation analysis, as follows:

Liu's criterion. If the characteristic equation of the positive equilibrium point is given by: $\lambda^3 + c_1(\alpha)\lambda^2 + c_2(\alpha)\lambda + c_3(\alpha) = 0$, where $c_1(\alpha)$, $c_2(\alpha)$ and $c_3(\alpha)$ are smooth functions of α in an open interval about $\alpha_c \in \mathcal{R}$ such that:

(a) $c_1(\alpha_c) > 0, \Delta(\alpha_c) = c_1(\alpha_c)c_2(\alpha_c) - c_3(\alpha_c) = 0, c_3(\alpha_c) > 0.$

(b) $\left(\frac{d\Delta}{d\alpha}\right)_{\alpha=\alpha_c} \neq 0,$

then a simple Hopf bifurcation occurs at $\alpha = \alpha_c$.

Applying the Liu's criterion to the characteristic equation (20), we observe that

$$c_1 = m_2 + m_3 + n_2 + \frac{rx_1^*}{K} > 0, \quad c_2 = (m_2 + m_3 + n_2)\frac{rx_1^*}{K} > 0, \quad c_3 = \alpha(m_2 + n_2)x_3^* > 0,$$

for all positive values of α .

Solving the equation $c_1(\alpha_c)c_2(\alpha_c) - c_3(\alpha_c) = 0$, we obtain

$$\alpha_c = \frac{rm_3(m_2 + n_2)}{\gamma_1 n_2 z K}, \tag{37}$$

where

$$z = -\frac{h_1}{2} + \sqrt{\frac{h_1^2}{4} + h_2},$$

$$h_1 = m_2 + m_3 + n_2 + \frac{m_3(m_2 + n_2)}{m_2 + m_3 + n_2},$$

$$h_2 = \frac{m_3(m_2 + n_2)(r - m_1 - n_1)}{m_2 + m_3 + n_2}.$$

Considering condition (b) of the *Liu's criterion*, we have

$$\left(\frac{d\Delta}{d\alpha}\right)_{\alpha=\alpha_c} = -\frac{B_1}{\alpha_c^2} - \frac{2B_2}{\alpha_c^3} < 0,$$

where

$$B_1 = \frac{rm_3(m_2 + n_2 + m_3)^2(m_2 + n_2) + rm_3(m_2 + n_2)^2}{\gamma_1 n_2 K},$$

$$B_2 = \frac{r^2(m_2 + n_2 + m_3)(m_2 + n_2)^2 + m_3^2}{\gamma_1^2 n_2^2 K^2}.$$

Hence, according to Liu's criterion, a simple Hopf bifurcation occurs at $\alpha = \alpha_c$, that is, $A_5 = 1$.

4 SEASONAL DYNAMICS OF THE SUGARCANE BORER-PARASITOID AGRO-ECOSYSTEM

Several environmental parameters (such as air temperature, air humidity, rainfall dispersion, among others) fluctuate periodically affecting an ecological system dynamics. Thus, they can be represented as periodic-time functions. In this section, the intrinsic growth rate r in system (1) is considered as a sinusoidal function representing these seasonal perturbations. The parameter r being defined by the following function Altizer et al. (2006); Gakkhar et al. (2003); Gakkhar and Naji (2003); Rinaldi et al. (1993):

$$r(t) = r_0 \left(1 + r_1 \sin \left(\frac{2\pi t}{365} \right) \right), \quad (38)$$

where t is measured in days, so r_0 is the average value of r over an integer number of years. The parameter r_1 represents the degree of seasonality, hence $r_0 r_1$ is the magnitude of the perturbation in r .

A population of pests in a crop can behave chaotically. Dennis et al. (2001) estimated the possibility of chaotic dynamics in insect populations, and the possibility of causing small cyclical chaotic events in the population dynamics of these organisms can be very useful for controlling these populations, assisting in pest management. In our study, chaotic dynamics occur with the inclusion of seasonality in the parameter r .

Next, we are interested in the seasonal dynamics of the equilibrium point E_5 ,

where there is coexistence of the parasitoid and pest populations. For this, we will keep the values of the parameters fixed as follows Parra et al. (2002); Rafikov and Holanda Limeira (2012); Rafikov and Silveira (2014).

$$\begin{aligned}
 m_1 = 0, \quad m_2 = 0.03566, \quad m_3 = \frac{1}{4}, \quad m_4 = 0.00257, \quad m_5 = m_4, \quad m_6 = \frac{1}{5}, \\
 n_1 = \frac{1}{8}, \quad n_2 = \frac{1}{9}, \quad n_3 = \frac{1}{50}, \quad n_4 = \frac{1}{16}, \quad r_0 = 0.19, \\
 \beta = 0.000009, \quad \gamma_1 = 2.29, \quad \gamma_2 = 40, \quad K = 25000.
 \end{aligned} \tag{39}$$

Setting the parameter values as specified in (39), it can be shown from (27) that without seasonality, that is, $r_1 = 0$ in (38), depending on the value of $\alpha > 0.169 \times 10^{-4}$, the attractor in the phase space can be an equilibrium point or a limit cycle, namely:

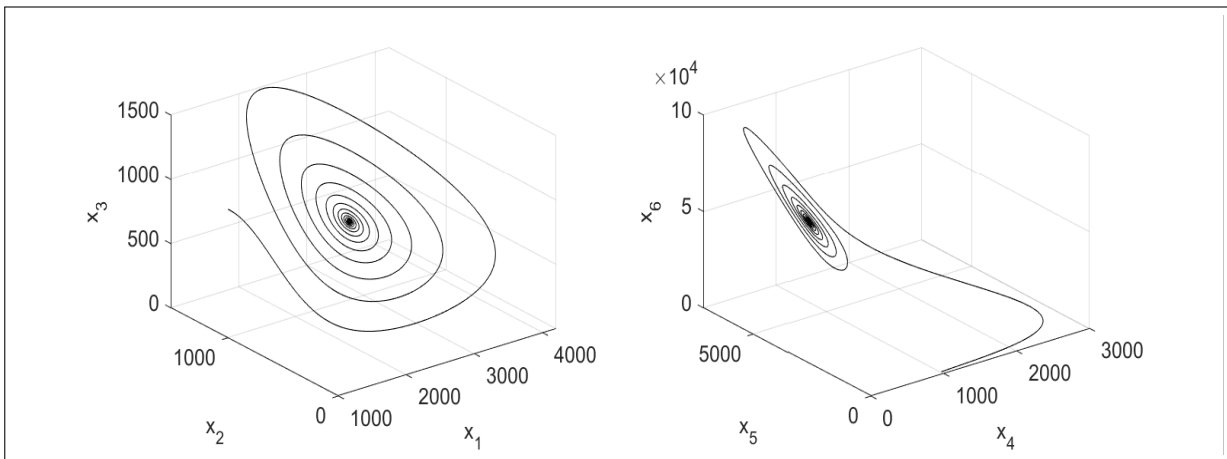
- For $0.169 \times 10^{-4} < \alpha < \alpha_c$, where $\alpha_c = 0.9135 \times 10^{-4}$, the corresponding attractor is the coexistence equilibrium point E_5 , whose components depend on the value of α , as specified in (26);
- For $\alpha > \alpha_c$, that is, beyond the Hopf bifurcation, the attractor is a period one limit-cycle and, the amplitude of this limit cycle in the phase space increases with increasing the value of α .

The addition of seasonality to the model, through the population growth rate according to (38), induces a destabilizing effect and may even trigger chaotic behavior. This destabilizing effect will be confirmed further by computer simulations with the parameter values fixed according to (39). The initial conditions are (1000,100,100,1000,100,300). In cases where there is chaos, the Lyapunov exponent was calculated, with simulations in the time interval $t \in [0, 3000]$. The figure caption shows the value around which the Lyapunov exponent oscillates.

First, we investigate the effect of seasonality for a value of α when the attractor is the equilibrium point E_5 in the 6D phase space, as shown in Fig. 1 for $\alpha = 0.6 \times 10^{-4}$. Considering this value of α , the bifurcation diagram for $0 \leq r_1 \leq 0.35$ is shown in Fig. 2, where it can be immediately verified that the value of $x_1(t)$ for $r_1 = 0$ is, as expected, the same value of this component of the equilibrium point obtained without seasonality, given by (26), and shown in Fig. 1. Increasing the value of r_1 , there is a periodic solution, which constitutes a route to chaos that occurs for $r_1 > 0.33$. The projections of the

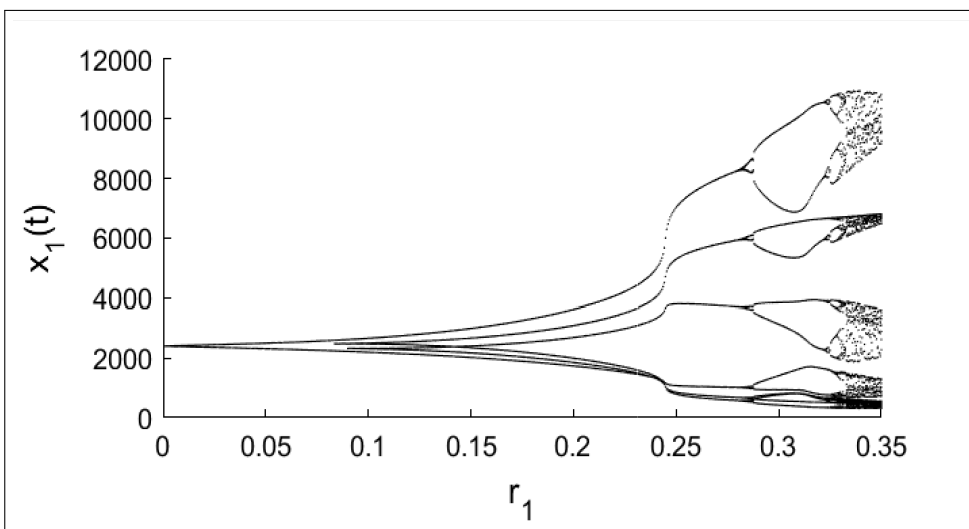
6D strange attractor, in the phase space, with $r_1 = 0.35$, are plotted in Fig. 3. Thus, the seasonality destabilizes the E_5 equilibrium, changing the attractor from equilibrium point to limit cycle, and then to more complex dynamics as r_1 increases.

Figure 1 – The equilibrium E_5 , reached by the populations of system (1), without seasonality, for the parameters fixed in (39) and $\alpha = 0.6 \times 10^{-4} < \alpha_c$



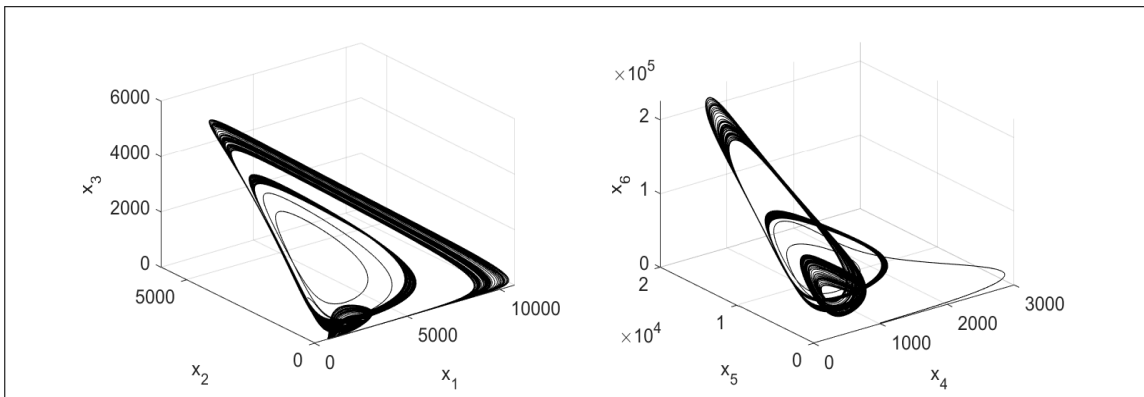
Source: the authors (2024)

Figure 2 – Keeping the parameter values fixed in (39) and $\alpha = 0.6 \times 10^{-4}$, the bifurcation diagram of $x_1(t)$ for $0 \leq r_1 \leq 0.35$



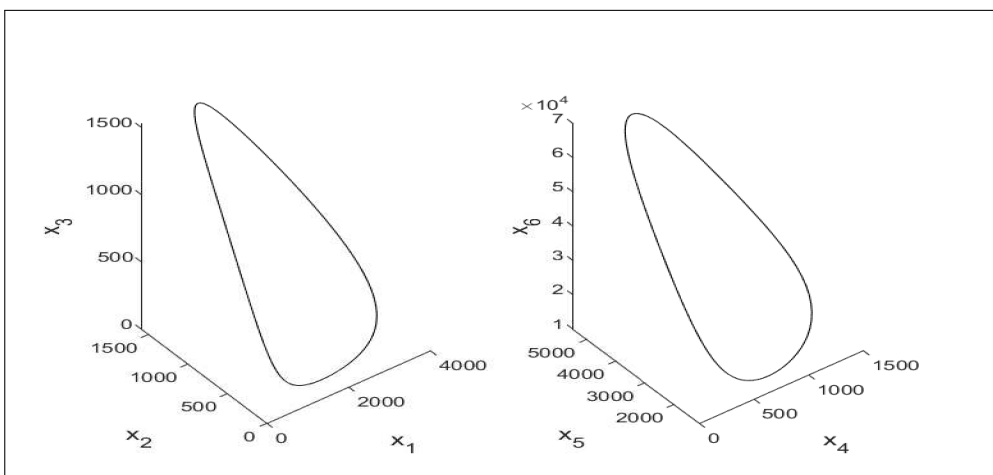
Source: the authors (2024)

Figure 3 – Keeping the parameter values fixed according to (39) and $\alpha = 0.6 \times 10^{-4}$, the projections of the 6D strange attractor in the phase space, corresponding to $r_1 = 0.35$, in the subspaces $x_1x_2x_3$ and $x_4x_5x_6$



Source: the authors (2024)

Figure 4 – The projection of the 6D limit cycle reached by the populations of system (1), without seasonality, for the parameters fixed in (39) and $\alpha = 1 \times 10^{-4} > \alpha_c$, in the subspaces $x_1x_2x_3$ and $x_4x_5x_6$



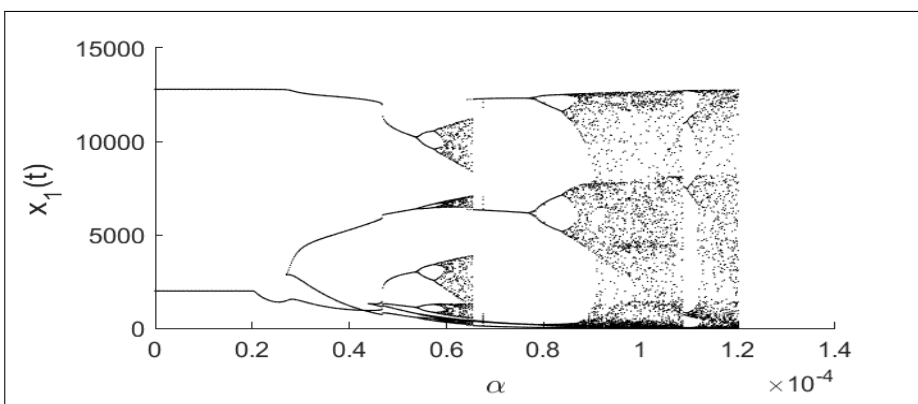
Source: the authors (2024)

Now, the range $\alpha > \alpha_c$ is considered, with the system having a period one limit cycle in the phase space, whose amplitude increases as α increases. We consider $\alpha = 1 \times 10^{-4}$ in Fig. 4, and the maximum and minimum values of x_1 as the same that as we identify for $r_1 = 0$ in the bifurcation diagram of $x_1(t)$ plotted with seasonality in Fig 6 for $0 \leq r_1 \leq 0.35$. Increasing the value of r_1 , a period doubling sequence can be noted, which constitutes a route to chaos that occurs for $r_1 > 0.2$. The projections of the 6D strange attractor in the phase space, corresponding to $r_1 = 0.25$, are plotted in Figure 5. Increasing even more the value of r_1 , periodic attractors emerge, as the one plotted in Fig. 7, corresponding to $r_1 = 0.28$, and similar periodic attractors occur

for $0.26 \leq r_1 \leq 0.3$. Therefore, the seasonality destabilizes the attractor from period one limit cycle, changing the behavior of our system to more complex dynamics as r_1 increases.

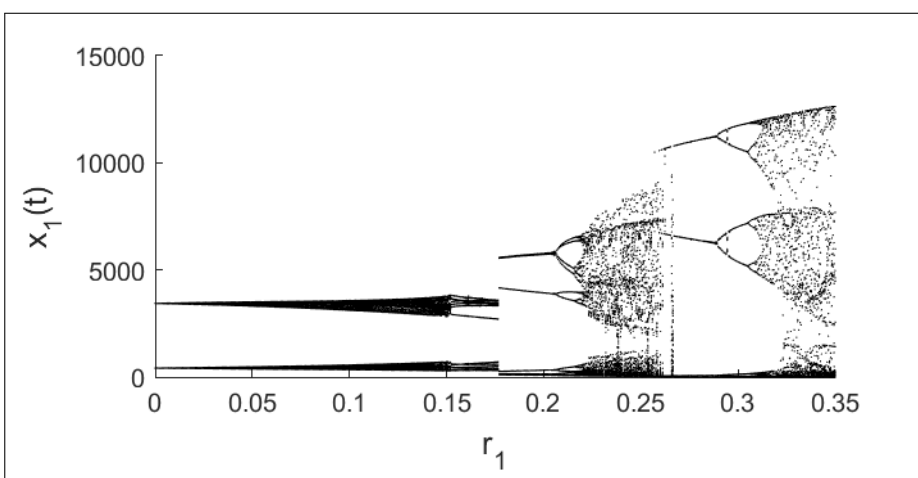
Regarding to the value of r_1 at which chaos is observed to occur, the comparison of bifurcation diagrams in Figs. 2 and 6 show that if $\alpha > \alpha_c$, chaos occurs at a lower value of r_1 than if $\alpha < \alpha_c$.

Figure 5 – Keeping the parameter values fixed in (39) and $\alpha = 1 \times 10^{-4}$, the projections of the 6D strange attractor in the phase space, corresponding to $r_1 = 0.25$, in the subspaces $x_1x_2x_3$ and $x_4x_5x_6$



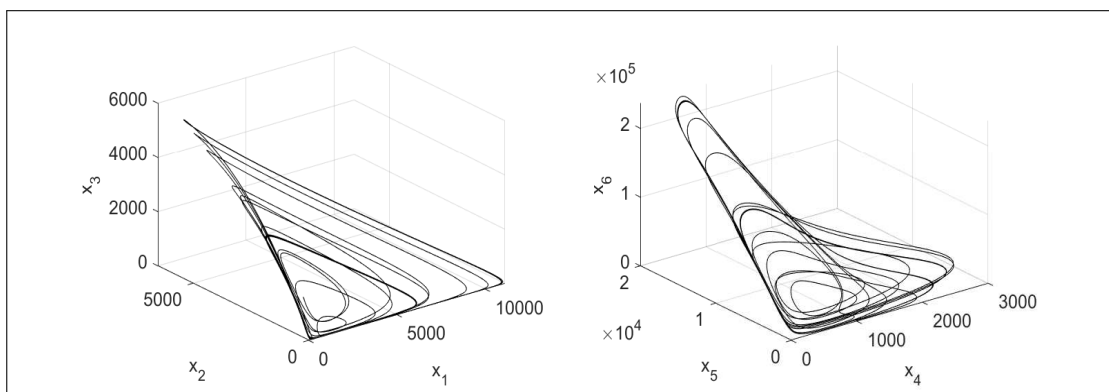
Source: the authors (2024)

Figure 6 – Keeping the parameter values fixed in (39) and $\alpha = 1 \times 10^{-4}$, the bifurcation diagram of $x_1(t)$ for $0 \leq r_1 \leq 0.35$



Source: the authors (2024)

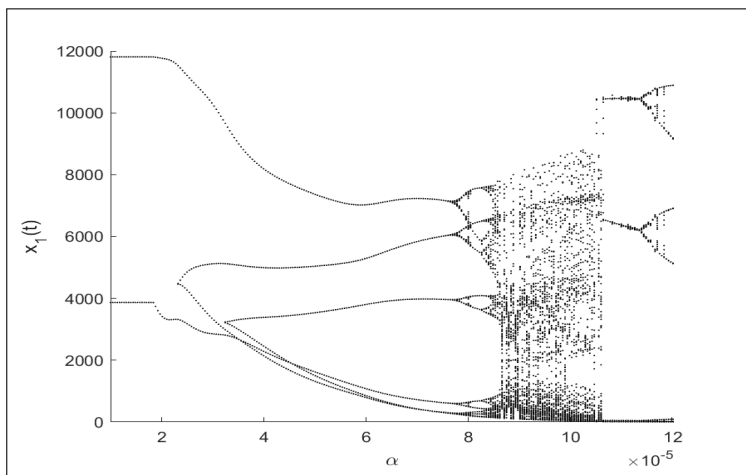
Figure 7 – Keeping the parameter values fixed in (39) and $\alpha = 1 \times 10^{-4}$, the projections of the 6D periodic attractor in the phase space, corresponding to $r_1 = 0.28$, in the subspaces $x_1x_2x_3$ and $x_4x_5x_6$



Source: the authors (2024)

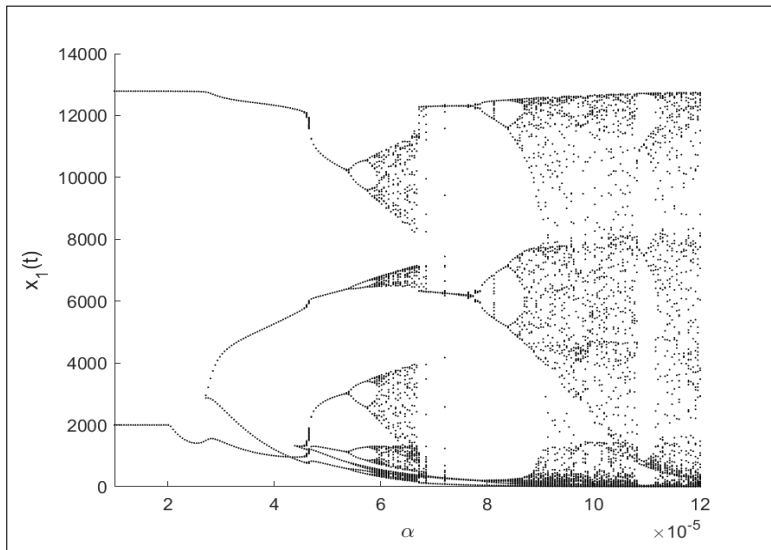
Additionally, we can investigate the effect of seasonality for fixed values of its degree r_1 , while varying the parameter α . For that, bifurcation diagrams of $x_1(t)$ are plotted in Figures 8 and 9, keeping the parameter values fixed in (39) for $0.169 \times 10^{-4} \leq \alpha \leq 1.2 \times 10^{-4}$, and setting values for the parameter r_1 . The diagram presents in Fig. 8 corresponds to $r_1 = 0.25$, whose strange attractor considering $\alpha = 1 \times 10^{-4}$ was visualized in Fig. 5, while the diagram in Fig. 9 the degree of seasonality is $r_1 = 0.35$, whose strange attractor for $\alpha = 0.6 \times 10^{-4}$ was visualized in Fig. 3. Comparing these two bifurcation diagrams, we conclude that the higher value of r_1 , the lower the value of α at which chaos is established, as it occurs at $\alpha = 0.8 \times 10^{-4}$ for $r_1 = 0.25$ and at $\alpha = 0.6 \times 10^{-4}$ for $r_1 = 0.35$.

Figure 8 – Bifurcation diagram of $x_1(t)$ for $0.169 \times 10^{-4} \leq \alpha \leq 1.2 \times 10^{-4}$, and keeping the parameter values fixed according to 39, for $r_1 = 0.25$



Source: the authors (2024)

Figure 9 – Bifurcation diagram of $x_1(t)$ for $0.169 \times 10^{-4} \leq \alpha \leq 1.2 \times 10^{-4}$, and keeping the parameter values fixed according to 39, for $r_1 = 0.35$



Source: the authors (2024)

5 CONCLUSIONS

According to White and Hastings (2020), seasonality is a significant feature in ecological systems driven by periodic climatic conditions, but it is often not explicitly included in either empirical or theoretical studies. Therefore this article is an effort toward the integration of the complex dynamics involving such seasonal influences into the sugarcane borer agroecosystem and its parasitoids. In this work, we have proposed a novel, six-dimensional continuous-time dynamical system, modeling interactions between the sugarcane borer and its egg and larval parasitoid. On the analytical side, five equilibrium states and conditions for their local stability were found out. Moreover, the Lyapunov function stability analysis ensured the global asymptotical stability of the equilibrium state in which all considered populations coexist. Then, the occurrence of a Hopf bifurcation was investigated applying Liu's theorem. Numerical simulations revealed the chaotic behavior of the system with seasonality. These results show how seasonality changes considerably the agroecosystem dynamics leading an asymptotically stable system, as shown in Fig 1 to the period-doubling and subsequently to a chaotic attractor shown in Fig 2. For a real system, this means sudden changes can occur in populations that without seasonality could coexist in an equilibrium. Moreover, when populations exhibit periodic

oscillations, as shown in Fig. 4, the introduction of seasonality can transform these oscillations into chaos, even for smaller values of r_1 than in the case of Fig. 6. Finally, bifurcation diagrams of the maximum and minimum population values, in the presence of seasonal influences, show that an increase in the parasitism coefficient α can lead the stabilized system to a chaotic regime, as shown by Figures 8 and 9.

The present results help understand the dynamics of the six-dimensional agro-ecological system with the seasonal forcing. Using these results the biological control strategies can be investigated in future research.

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Contribution: Conceptualization; Data curation; Formal Analysis; Investigation; Methodology; Supervision; Validation; Writing – review; Editing

How to cite this article

Rafikov, M., Molter, A., Bezerra, J. I. M., Rafikova, E., & Varriale, M. C. (2024). Chaotic dynamics of the sugarcane borer-two parasitoid agroecosystem with seasonality. *Ciência e Natura*, Santa Maria, v.46, e75328. DOI: <https://doi.org/10.5902/2179460X75328>.