



Article **Fractional Order Modeling the Gemini Virus in** *Capsicum annuum* with Optimal Control

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Abstract: In this article, a fractional model of the *Capsicum annuum* (*C. annuum*) affected by the yellow virus through whiteflies (*Bemisia tabaci*) is examined. We analyzed the model by equilibrium points, reproductive number, and local and global stability. The optimal control methods are discussed to decrease the infectious *B. tabaci* and *C. annuum* by applying the *Verticillium lecanii* (*V. lecanii*) with the Atangana–Baleanu derivative. Numerical results described the population of plants and comparison values of using *V. lecanni*. The results show that using 60% of *V. lecanni* will control the spread of the yellow virus in infected *B. tabaci* and *C. annuum* in 10 days, which helps farmers to afford the costs of cultivating chili plants.

Keywords: fractional calculus; stability analysis; Atangana–Baleanu fractional derivative; optimal control; mathematical models; yellow virus

MSC: 34A08; 26A33; 03C50; 35F21

1. Introduction

Growth of *Capsicum annuum* (chili plant) [1,2] is excessive in the mid-hill region of India. It contains vitamin C, provitamin A, and calcium, which are good for health. The spicy taste and high nutritional benefit the marketing of *C. annuum*. *Capsicum annuum* is used in the pharmaceutical industries to increase immunity, antiulcer, analgesic, antidiabetic, and antihemorrhoid agents. The extracts of *C. annuum* employ to relieve the pain of inflammation of joints, headaches, neuralgia, and burns. The framers require advantages to yield the *C. annuum* in large amounts. Due to natural obstacles like soil erosion, irrigation, and diseases spread, farmers encounter heavy losses while fertilizing *C. annuum*.

The cause of Geminivirus (yellow virus) [3–5] is one of the difficulties experienced by farmers in the cultivation of *C. annuum*. Yellow spots appear in young leaves and shoots, and the leaves turn out to be bright yellow or mixed yellow-green, which are symptoms of the yellow virus. This virus spreads by whitefly (*Bemisia tabaci*) from one host to another continuously.

Controlling these viruses using overlapping cropping methods is very difficult, since insecticide must be applied to mature plants. Controlling techniques vary depending on the conditions of the plants infected by virus variety, environment, and time. To reduce the populations of white-flies, systemic insecticides are applied to control the spread of the virus, as well as to cure the infected white-fly insect and plants, like rust fungi, etc., who have the host cyclodepsipeptide toxin. This toxin was produced via the mycelium of entomopathogenic fungi (*Verticillium lecanii*) [6–8]. However, the excess use of *V. lecanii*



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Copyright: © 2022 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https:// creativecommons.org/licenses/by/ 4.0/). generates high costs. To minimize costs of controlling the *B. tabaci* population, an optimal control method must be found.

Over the past few decades, numerous analyses, real-world problems, and numerical methods were resolved by fractional derivatives and integrals [9–16]. The applications are fluid mechanics, electrochemistry, viscoelasticity, optics, and signals processing in engineering and science. Riemann–Liouville, Caputo, Caputo–Fabrizio, and Atangana–Baleanu are some of the fractional derivatives developed by several researchers. The Atangana–Baleanu fractional derivative (AB-derivative) [17] is a new one among the Mittag–Leffler Kernel. Recently, the existence analysis [18–21], stability analysis [22,23] and models [24–27] of AB-derivative were elaborated by many authors.

Optimal control is the approach of ascertaining control and circumstances path for dynamic systems to minimize an accomplishment period. The origin of the optimal control is related to the calculus of variations. In the 1940s, the formulation of dynamic programming in the optimal control was developed by Rochard Bellman. Using the analytical method, some of the optimal control problems' solutions are difficult to find. N.H. Sweilam et al. [28] discussed the optimal control method for cancer treatment using AB-derivative. R. Amelia et al. [29] showed results to help farmers afford the costs of cultivating the red chilies by optimal control. In [30] N.H. Sweilam and S.M. AL–Mekhlafi described the fractional model of multistrain TB cure with optimal control. The optimal control problems to solve numerical procedures were investigated in [31,32].

To the best of our knowledge, the study of the *C. annuum* of the yellow virus with optimal control by applying the AB-derivative to the model is yet to come. This article was organized as follows: the basic results and definitions of the AB-derivative are discussed in Section 2. In Section 3, the formation of the *C. annuum* model with AB-derivative is explained. In Section 4, the optimality conditions demonstrate. In Sections 5 and 6, numerical results with graphs for the fractional optimal control problem have presented and conclusions.

Motivated by [22,23,28,30], this document discusses the fractional model of Geminivirus in *C. annuum* with AB-derivative via optimal control and stability analysis. The main contributions are organized as follows:

- (A) The fractional model of Geminivirus in *C. annuum* with AB-derivative constructed.
- (B) We obtained some stability results of this fractional model and discussed the equilibrium points and reproductive number of the model.
- (C) We derived the optimal control of this fractional model and plotted the population and comparison results of each variable in the model.

2. Preliminaries

This section briefly discussed some preliminaries regarding fractional derivatives. There are few definitions for the fractional derivatives, including Riemann–Liouville, Caputo, and Caputo–Fabrizio [9,18]. Recently, a new fractional derivative with Mittag–Leffler Kernel was elaborated and implemented in a few real-world models [24–27]. We present the following definitions.

The Riemann–Liouville fractional integral (RL) is defined as follows [9,18]

$$(_0I^{\zeta}\phi)(t) = \frac{1}{\Gamma(\zeta)}\int_0^t (t-s)^{\zeta-1}\phi(s)ds, \ \zeta > 0.$$

The Riemann–Liouville fractional order derivative (RL) is defined as follows [9,18]

$$\binom{R}{0}D^{\zeta}\phi(t) = \frac{d}{dt}\left(\frac{1}{\Gamma(1-\zeta)}\int_0^t (t-s)^{-\zeta}\phi(s)ds\right), \ 0 < \zeta < 1.$$

The Caputo fractional order derivative (C) is defined as follows [9,18]

$$({}_{0}^{C}D^{\zeta}\phi)(t) = rac{1}{\Gamma(1-\zeta)}\int_{0}^{t}(t-s)^{-\zeta}\phi'(s)ds, \ 0 < \zeta < 1.$$

The Caputo-Fabrizio fractional order derivative in Caputo sense (CFC) is defined as follows [33]

$$\binom{CFC}{0}D^{\zeta}\phi(t) = \frac{M(\zeta)}{1-\zeta}\int_0^t \phi'(s)exp\bigg[\frac{-\zeta}{1-\zeta}(t-s)^{\zeta}\bigg]ds, \ 0<\zeta<1.$$

where $\phi' \in H^1(0, T)$, $M(\zeta)$ is a constant of normalization that depends on ζ , which satisfies that, M(0) = M(1) = 1.

The Atangana–Baleanu fractional order derivative in the Riemann–Liouville sense (ABR) is defined as follows [17]

$$\binom{ABR}{0}D^{\zeta}\phi(t) = \frac{B(\zeta)}{1-\zeta}\frac{d}{dt}\int_0^t\phi(s)E_{\zeta}\left[-\zeta\frac{(t-s)^{\zeta}}{1-\zeta}\right]ds, \ 0<\zeta<1.$$

where $\phi' \in H^1(0, T)$, $B(\zeta)$ is a normalization function, B(0) = B(1) = 1.

The Atangana–Baleanu fractional order derivative in the Caputo sense (ABC) is defined as follows [17]

$$\binom{ABC}{0}D^{\zeta}\phi(t) = \frac{B(\zeta)}{1-\zeta}\int_0^t \phi'(s)E_{\zeta}\left[-\zeta\frac{(t-s)^{\zeta}}{1-\zeta}\right]ds, \ 0<\zeta<1.$$

where $\phi' \in H^1(0, T)$, $B(\zeta)$ is a normalization function, B(0) = B(1) = 1.

The Atangana–Baleanu fractional integral of order ζ of a function $\phi(t)$ is defined as [17]

$$({}^{AB}_0I^{\zeta}\phi)(t)=\frac{1-\zeta}{B(\zeta)}\phi(t)+\frac{\zeta}{B(\zeta)}({}_0I^{\zeta}\phi)(t).$$

The Mittag–Leffler function of one and two parameters $E_{\alpha}(z)$, $E_{\alpha,\beta}(z)$ is defined as [9]

$$E_{\alpha}(z) = \sum_{k=0}^{\infty} \frac{z^{k}}{\Gamma(\alpha k+1)} z, \alpha \in C, Re(\alpha) > 0.$$
$$E_{\alpha,\beta}(z) = \sum_{k=0}^{\infty} \frac{z^{k}}{\Gamma(\alpha k+\beta)} z, \alpha, \beta \in C, Re(\alpha) > 0, Re(\beta) > 0$$

The generalized Mittag-Leffler function is defined as [9]

$$E_{\alpha,\beta}^{\gamma}(z) = \sum_{k=0}^{\infty} \frac{(\gamma)_n}{\Gamma(\alpha k + \beta)} \frac{z^k}{k!} z, \alpha, \beta, \gamma \in C, Re(\alpha) > 0, Re(\beta) > 0, Re(\gamma) > 0,$$

where $\Gamma(\cdot)$ denotes the Gamma function, and note that

$$E_{1,1}^{1}(z) = e^{z}, E_{\alpha,1}^{1}(z) = E_{\alpha}(z), E_{\alpha,\beta}^{1}(z) = E_{\alpha,\beta}(z)$$

3. Modeling Framework of Gemini Virus

The fractional model based on the cure of yellow virus in *C. annuum* by *V. lecanii* with modified variables and parameters is presented. Here, the parameters depend on the

fractional model, which is an extension of the integer model given in [29]. The mathematical model of *C. annuum* with AB fractional derivative is represented as follows:

$$\begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (S_{v}(t)) = A - \alpha S_{v} - \beta_{1}(1 - \delta_{p})S_{v}I_{BT} - \mu_{p}S_{v} \\ \begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (I_{v}(t)) = \beta_{1}(1 - \delta_{p})S_{v}I_{BT} - \mu_{p}I_{v} \\ \begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (S_{g}(t)) = \alpha S_{v} - \beta_{2}(1 - \delta_{p})S_{g}I_{BT} - \mu_{p}S_{g} \\ \begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (I_{g}(t)) = \beta_{2}(1 - \delta_{p})S_{g}I_{BT} - \mu_{p}I_{g} \\ \begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (S_{BT}(t)) = BN_{v} - \gamma_{1}(1 - \delta_{p})I_{v}S_{BT} - \gamma_{2}(1 - \delta_{p})I_{g}S_{BT} \\ - \theta_{1}\delta_{p}S_{BT}N_{p} - \mu_{1}S_{BT} \\ \begin{pmatrix} A^{BC}D_{t}^{\zeta} \\ 0 \end{pmatrix} (I_{BT}(t)) = \gamma_{1}(1 - \delta_{p})I_{v}S_{BT} + \gamma_{2}(1 - \delta_{p})I_{g}S_{BT} \\ - \theta_{1}\delta_{p}I_{BT}N_{p} - \mu_{1}I_{BT} \\ \text{with } S_{v}(0) = S_{v0}, I_{v}(0) = I_{v0}, \\ S_{g}(0) = S_{g0}, I_{g}(0) = I_{g0}, \\ S_{BT}(0) = S_{BT0}, I_{BT}(0) = I_{BT0}, \end{cases}$$

$$(1)$$

where $0 < \zeta < 1$.

The total population is denoted by N_p of *C. annuum* $N_p = S_v + I_v + S_g + I_g$ is taken to be constant. The total population of *B. tabaci* is denoted by $N_v = S_{BT} + I_{BT}$. Here, the total population can be divided into 6 subgroups.

- S_v denotes a set of noninfected *C. annuum* in vegetative phase liable to possible infection.
- *I_v* denotes a set of infected *C. annuum* in vegetative phase.
- S_g denotes a set of noninfected *C. annuum* in generative phase liable to possible infection.
- *I_g* denotes a set of infected *C. annuum* in generative phase.
- *S*_{BT} denotes a set of noninfected *B. tabaci* (white bug) liable to possible infection.
- I_{BT} denotes a set of infected *B. tabaci*.

The recruitment rate of *C. annuum* and *B. tabaci* is denoted by *A* and *B* respectively. The growth rate of *C. annuum* from vegetative to generative phase is denoted by α . β_1 , and β_2 denoted the infection rate of *C. annuum* in the vegetative and generative phase respectively. γ_1 , and γ_2 denoted the infection rate of *B. tabaci* in the vegetative and generative phase respectively. δ_p stands for the rate of use of *V. lecanii*. The death rate of *C. annuum* is denoted by μ_p . The natural death rate of *B. tabaci* is denoted by μ_1 , and the death rate of *B. tabaci* due to curative intervention is denoted by θ_1 .

4. Basic Analysis of the Model

4.1. Invariant Region

The fractional order *C. annuum* model of yellow virus (1) can be analyzed in the biological feasible region discussed as follows. The system (1) is split into two parts, namely the *C. annuum* population (N_p ; with $N_p = S_v + I_v + S_g + I_g$) and the *B. tabaci* population (N_v ; with $N_v = S_{BT} + I_{BT}$).

Let the feasible region $F = F_p \cup F_v \subset R^4_+ \times R^2_+$ with

$$F_{p} = \left\{ (S_{v}, I_{v}, S_{g}, I_{g}) \in R_{+}^{4} : S_{v} + I_{v} + S_{g} + I_{g} \leq \frac{A}{\mu_{p}} \right\},\$$

$$F_{v} = \left\{ (S_{BT}, I_{BT}) \in R_{+}^{2} : S_{BT} + I_{BT} \leq \frac{B}{\theta_{1}\delta_{p}N_{p} + \mu_{1}} \right\}.$$

To establish the positive invariance of $F\langle i.e., solutions in F remain in F for all <math>t > 0 \rangle$. Adding the first four equations and the last two equations of the model (1) gives

$${}_{0}^{ABC}D_{t}^{\varsigma}N_{p}(t) = A - \mu_{p}(S_{v} + I_{v} + S_{g} + I_{g}) = A - \mu_{p}N_{p}$$
(2)

$${}_{0}^{ABC}D_{t}^{\zeta}N_{v}(t) = BN_{v} - (\theta_{1}\delta_{p}N_{p} + \mu_{1})N_{v} = B - (\theta_{1}\delta_{p}N_{p} + \mu_{1})N_{v}$$
(3)

This can be used to show that the fractional order of the *C. annuum* and *B. tabaci* population in the system (1) shows that

$$A_0^{BC} D_t^{\zeta} N_p(t) \le A - \mu_p N_p,$$

$$A_0^{BC} D_t^{\zeta} N_v(t) \le B - (\theta_1 \delta_p N_p + \mu_1) N_v.$$

$$\left.\right\}$$

$$(4)$$

which implies that

$$N_{p}(t) \leq At^{\zeta} E_{\zeta,\zeta+1}(-\mu_{p}t^{\zeta}) - N_{p}(0)E_{\zeta,1}(-\mu_{p}t^{\zeta}),$$

$$N_{v}(t) \leq Bt^{\zeta} E_{\zeta,\zeta+1}(-(\theta_{1}\delta_{p}N_{p}+\mu_{1})t^{\zeta}) + N_{v}(0)E_{\zeta,1}(-(\theta_{1}\delta_{p}N_{p}+\mu_{1})t^{\zeta}).$$

From above inequality, we observe that $N_p(t) \leq \frac{A}{\mu_p} \& N_v(t) \leq \frac{B}{(\theta_1 \delta_p N_p + \mu_1)}$. Thus, the region *F* is positively-invariant.

Hence, it is sufficient to consider the dynamics model of system (1) in F. The mathematical model is well-posed in the region F.

: Every solution of the basic model (1) with initial conditions in *F* remains in *F* for all t > 0. The result is summarized below.

Lemma 1. The region $F = F_p \cup F_v \subset R^4_+ \times R^2_+$ is positively invariant for the basic model (1) with non-negative initial conditions in R^6_+ .

4.2. Disease-Free Equilibrium Point

To evaluate the equilibrium points Let

$${}^{ABC}_{0}D^{\zeta}_{t}S_{v}(t) = 0, {}^{ABC}_{0}D^{\zeta}_{t}S_{g}(t) = 0, {}^{ABC}_{0}D^{\zeta}_{t}S_{BT}(t) = 0,$$

$${}^{ABC}_{0}D^{\zeta}_{t}I_{v}(t) = 0, {}^{ABC}_{0}D^{\zeta}_{t}I_{g}(t) = 0, {}^{ABC}_{0}D^{\zeta}_{t}I_{BT}(t) = 0.$$

Then

$$\mathcal{E}_{0}(S_{v}^{0}, I_{v}^{0}, S_{g}^{0}, I_{g}^{0}, S_{BT}^{0}, I_{BT}^{0}) = \left(\frac{A}{\mu_{p}}, 0, 0, 0, \frac{B}{\theta_{1}\delta_{p}N_{p} + \mu_{1}}, 0\right).$$

4.3. Reproduction Number

For the basic reproduction number for the *C. annuum* model (1), suppose that $y = (S_v, S_g, I_{BT})$ and using next generation matrix approach [34], we have

$$\frac{dy}{dt} = \mathcal{F}(y) - \mathcal{V}(y),$$

where Jacobian of \mathcal{F} and \mathcal{V} at \mathcal{E}_0 , we have

$$\mathcal{F} = \begin{pmatrix} \alpha + \mu_p & 0 & 0 \\ -\alpha & \mu_p & 0 \\ 0 & 0 & \mu_1 \end{pmatrix} \quad \& \mathcal{V} = \begin{pmatrix} \frac{\beta_1(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & 0 & \frac{\beta_1(1-\delta_p)A}{\mu_p} \\ 0 & \frac{\beta_2(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & \frac{\beta_2(1-\delta_p)A}{\mu_p} \\ 0 & 0 & \theta_1\delta_pN_p \end{pmatrix}.$$

The basic reproduction number ψ_0 comes from the spectral radius $\psi_0 = \rho[\mathcal{FV}^{-1}]$, given by

$$\psi_0 = \frac{(\alpha + \mu_p)(\theta_1 \delta_p N_p + \mu_1)}{\beta_1 (1 - \delta_p) B},$$

where

$$\mathcal{FV}^{-1} = \begin{pmatrix} \frac{(\alpha + \mu_p)(\theta_1 \delta_p N_p + \mu_1)}{\beta_1 (1 - \delta_p) B} & 0 & 0\\ 0 & \frac{\mu_p(\theta_1 \delta_p N_p + \mu_1)}{\beta_2 (1 - \delta_p) B} & 0\\ 0 & 0 & \frac{\mu_1}{\theta_1 \delta_p N_p} \end{pmatrix}.$$

Theorem 1. There exists a unique positive endemic equilibrium point \mathcal{E}^* for system (1) if $\psi_0 > 1$.

Proof. Endemic equilibrium point is obtained from system (1), and by putting right-hand side of each equation equal to zero, we have

$$S_v^* = \frac{A}{a}$$

$$I_v^* = \frac{A\beta_1(1-\delta_p)I_{BT}}{\mu_p a}$$

$$S_g^* = \frac{A\alpha}{ab}$$

$$I_g^* = \frac{A\alpha\beta_2(1-\delta_p)I_{BT}}{\mu_p ab}$$

$$S_{BT}^* = \frac{BN_v\mu_p ab}{\gamma_1(1-\delta_p^2A\beta_1)I_{BT}b + \gamma_2\beta_2\alpha A(1-\delta_p)^2I_{BT} + (\theta_1\delta_pN_p - \mu_1)\mu_p ab'}$$

where $a = \alpha + \beta_1(1 - \delta_p)I_{BT} - \mu_p$ and $b = \beta_2(1 - \delta_p)I_{BT} + \mu_p$ and I_{BT}^* is the positive root of $\mathcal{K}(I_{BT}) = i_5 I_{BT}^5 + i_4 I_{BT}^4 + i_3 I_{BT}^3 + i_2 I_{BT}^2 + i_1 I_{BT} = 0$, where

$$\begin{split} i_{1} &= (\theta_{1}\delta_{p}N_{p} + \mu_{1})^{2}\mu_{p}^{4}(\alpha - \mu_{p})[(\alpha - \mu_{p}) + 2\beta_{1}(1 - \delta_{p})] \\ &- ABN_{v}(\gamma_{1}\beta_{1}\mu_{p}^{2} + \alpha\gamma_{2}\beta_{2}\mu_{p})(1 - \delta_{p})^{2}[\alpha\mu_{p} - \mu_{p}^{2}], \\ i_{2} &= (A\alpha\gamma_{1}\beta_{1}\mu_{p}^{2}(1 - \delta_{p})^{2} - A\gamma_{1}\beta_{1}\mu_{p}^{3}(1 - \delta_{p})^{2} + A\alpha^{2}\gamma_{2}\beta_{2}\mu_{p}^{2}(1 - \delta_{p})^{2} - A\alpha\gamma_{2}\beta_{2}\mu_{p}^{3}(1 - \delta_{p})^{2} \\ &+ 2\beta_{2}(\theta_{1}\delta_{p}N_{p} + \mu_{1})\mu_{p}^{3}(\alpha - \mu_{p})^{2}(1 - \delta_{p}))(\theta_{1}\delta_{p}N_{p} + \mu_{1}) - ABN_{v}\gamma_{1}\beta_{1}\beta_{2}\mu_{p}(1 - \delta_{p})^{3}[\alpha\mu_{p} - \mu_{p}^{2}] \\ &- ABN_{v}(\gamma_{1}\beta_{1}\mu_{p}^{2} + \alpha\gamma_{2}\beta_{2}\mu_{p})(1 - \delta_{p})^{2}[\alpha\beta_{2}(1 - \delta_{p}) + \mu_{p}\beta_{1}(1 - \delta_{p}) - \mu_{p}\beta_{2}(1 - \delta_{p})], \\ i_{3} &= (A\alpha\gamma_{1}\beta_{1}\beta_{2}\mu_{p}(1 - \delta_{p})^{3} + A\gamma_{1}\beta_{1}^{2}\mu_{p}(1 - \delta_{p})^{3} - A\gamma_{1}\beta_{1}\beta_{2}\mu_{p}^{2}(1 - \delta_{p})^{3} + A\alpha^{2}\gamma_{2}\beta_{2}^{2}\mu_{p}(1 - \delta_{p})^{3} \\ &+ (\theta_{1}\delta_{p}N_{p} + \mu_{1})\mu_{p}^{2}\Big[\beta_{2}^{2}(\alpha - \mu_{p})^{2}(1 - \delta_{p})^{2} + 4\beta_{1}\beta_{2}\mu_{p}(1 - \delta_{p})^{2}(\alpha - \mu_{p}) + \beta_{1}^{2}\mu_{p}^{2}(1 - \delta_{p})^{2}\Big]) \\ &\times (\theta_{1}\delta_{p}N_{p} + \mu_{1}) - ABN_{v}\gamma_{1}\beta_{1}\beta_{2}\mu_{p}(1 - \delta_{p})^{3}[\alpha\beta_{2}(1 - \delta_{p}) + \mu_{p}\beta_{1}(1 - \delta_{p}) - \mu_{p}\beta_{2}(1 - \delta_{p})] \\ &- ABN_{v}\gamma_{1}\beta_{1}^{2}\beta_{2}\mu_{p}^{2}(1 - \delta_{p})^{4} - ABN_{v}\alpha\gamma_{2}\beta_{1}\beta_{2}^{2}\mu_{p}(1 - \delta_{p})^{4}, \\ i_{4} &= (A\gamma_{1}\beta_{1}^{2}\beta_{2}\mu_{p}(1 - \delta_{p})^{4} + A\alpha\gamma_{2}\beta_{1}\beta_{2}^{2}(\alpha - \mu_{p}) + 2\mu_{p}\beta_{1}^{2}\beta_{2}(1 - \delta_{p})^{3}])(\theta_{1}\delta_{p}N_{p} + \mu_{1}) \\ &- ABN_{v}\gamma_{1}\beta_{1}^{2}\beta_{2}\mu_{p}(1 - \delta_{p})^{5}, \\ i_{5} &= \beta_{1}^{2}\beta_{2}^{2}(1 - \delta_{p})^{4}(\theta_{1}\delta_{p}N_{p} + \mu_{1})^{2}\mu_{p}^{2}. \end{split}$$

It is obvious from the values of $S_v^*, I_v^*, S_g^*, I_g^*, S_{BT}^* \& I_{BT}^*$ that there exists a unique positive endemic equilibrium point \mathcal{E}^* , if $\psi(0) > 1$. \Box

Theorem 2. The system (1) is locally stable at \mathcal{E}_0 for $\psi_0 < 1$ and unstable for $\psi_0 > 1$.

Proof. The Jacobian of system (1) is

$$J = \begin{pmatrix} Q_1 & 0 & 0 & 0 & \frac{-\beta_1(1-\delta_p)A}{\mu_p} \\ \frac{\beta_1(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & -\mu_p & 0 & 0 & 0 & \frac{\beta_1(1-\delta_p)A}{\mu_p} \\ \alpha & 0 & Q_2 & 0 & 0 & \frac{-\beta_2(1-\delta_p)A}{\mu_p} \\ 0 & 0 & \frac{\beta_2(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & -\mu_p & 0 & \frac{\beta_2(1-\delta_p)A}{\mu_p} \\ 0 & \frac{-\gamma_1(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & 0 & \frac{-\gamma_2(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & Q_3 & 0 \\ 0 & \frac{\gamma_1(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & 0 & \frac{\gamma_2(1-\delta_p)B}{\theta_1\delta_pN_p+\mu_1} & Q_4 & Q_5 \end{pmatrix}$$

where

$$\begin{aligned} Q_1 &= -[\alpha + \mu_p + \beta_1 (1 - \delta_p) I_{BT}], \\ Q_2 &= -[\mu_p + \beta_2 (1 - \delta_p) I_{BT}], \\ Q_3 &= -[\gamma_1 (1 - \delta_p) I_v + \gamma_2 (1 - \delta_p) I_g + \theta_1 \delta_p N_p + \mu_1], \\ Q_4 &= \gamma_1 (1 - \delta_p) I_v + \gamma_2 (1 - \delta_p) I_g, \\ Q_5 &= -[\theta_1 \delta_p N_p + \mu_1]. \end{aligned}$$

Along \mathcal{E}_0 , it implies that

$$J(\mathcal{E}_0) = \begin{pmatrix} -(\alpha + \mu_p) & 0 & 0 & 0 & 0 & 0 \\ 0 & -\mu_p & 0 & 0 & 0 & 0 \\ \alpha & 0 & -\mu_p & 0 & 0 & 0 \\ 0 & 0 & 0 & -\mu_p & 0 & 0 \\ 0 & 0 & 0 & 0 & -(\theta_1 \delta_p N_p + \mu_1) & 0 \\ 0 & 0 & 0 & 0 & 0 & -(\theta_1 \delta_p N_p + \mu_1) \end{pmatrix}$$

which follows that all the eigenvalues are negative if $\psi_0 < 1$ and eigenvalues are positive for $\psi_0 > 1$. Hence, we conclude that the system (1) is locally stable under the condition $\psi_0 < 1$ and unstable for $\psi_0 > 1$. \Box

Theorem 3. *The system* (1) *is globally stable, if* $\psi_0 > 1$ *at* \mathcal{E}_0 *.*

Proof. First, we construct the Lyapunov function $\mathcal{L}(t)$, for the system as:

$$\mathcal{L}(t) = 1 + I_{BT}(t) - \frac{ln I_{BT}(t)}{I_{BT}(0)}.$$
(5)

Then, differentiating the Equation (5) with respect to time, we have

$$\begin{aligned} \frac{d}{dt}(\mathcal{L}(t)) &= \frac{dI_{BT}(t)}{dt} - \frac{1}{I_{BT}(t)} \frac{dI_{BT}(t)}{dt} \\ &= \left(1 - \frac{1}{I_{BT}}\right) \frac{dI_{BT}}{dt} \\ &= \frac{dI_{BT}}{dt} - \theta_1 \delta_p N_p - \mu_1. \end{aligned}$$

By manipulating along the point \mathcal{E}_0 , we get

$$\frac{d}{dt}(\mathcal{L}(t)) = -(\theta_1 \delta_p N_p + \mu_1)$$

 $\leq 0 \text{ for } \psi_0 > 1.$

Again differentiating the above equation, we have

$$\frac{d^2}{dt}(\mathcal{L}(t)) = \frac{d^2 I_{BT}(t)}{dt} - \frac{dN_p}{dt}$$
$$= \frac{d^2 I_{BT}(t)}{dt} - 4\mu_p.$$

By manipulating along the point \mathcal{E}_0 , we get

$$\begin{aligned} \frac{d^2}{dt}(\mathcal{L}(t)) &= -4\mu_p \\ &\leq 0 \text{ for } \psi_0 > 1. \end{aligned}$$

Therefore, if $\psi_0 > 1$, then $\frac{d}{dt}(\mathcal{L}(t)) < 0$, which implies that the system (1) is globally stable for $\psi_0 > 1$ at \mathcal{E}_0 . \Box

Remark 1. In the case of $\psi_0 < 1$ at \mathcal{E}^* , it is an interesting problem to find an effective strategy to prevent the disease.

5. Optimal Control

The purpose of the dynamic red chili model is to minimize the population of plants infected during vegetative or generative period and insects infected by optimizing *V. lecanni* using AB-derivative [28–30].

The objective functions used are as follows:

$$J(u) = \int_0^{T_f} (A_1 I_v(t) + A_2 I_g(t) + A_3 I_{BT} + A_4 u_1^2(t)) dt,$$
(6)

where u_1 is the late of giving *V*. *lecanni* and $A_i \ge 0$, for i = 1, 2, ..., 4 is the cost coefficient and t_f is end time in $[0, T_f]$.

Therefore, by using u_1 *V. lecanni* in the Equation (1), it becomes

$$\begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (S_v(t)) = A - \alpha S_v - \beta_1 (1 - u_1) S_v I_{BT} - \mu_p S_v \\ \begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (I_v(t)) = \beta_1 (1 - u_1) S_v I_{BT} - \mu_p I_v \\ \begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (S_g(t)) = \alpha S_v - \beta_2 (1 - u_1) S_g I_{BT} - \mu_p S_g \\ \begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (I_g(t)) = \beta_2 (1 - u_1) S_g I_{BT} - \mu_p I_g \\ \begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (S_{BT}(t)) = B N_v - \gamma_1 (1 - u_1) I_v S_{BT} - \gamma_2 (1 - u_1) I_g S_{BT} \\ - \theta_1 u_1 S_{BT} N_p - \mu_1 S_{BT} \\ \begin{pmatrix} {}^{ABC}D_t^{\zeta} \\ 0 \end{pmatrix} (I_{BT}(t)) = \gamma_1 (1 - u_1) I_v S_{BT} + \gamma_2 (1 - u_1) I_g S_{BT} \\ - \theta_1 u_1 I_{BT} N_p - \mu_1 I_{BT} \\ \text{with } S_v(0) = S_{v0}, I_v(0) = I_{v0}, \\ S_g(0) = S_{g0}, I_g(0) = I_{g0}, \\ S_{BT}(0) = S_{BT0}, I_{BT}(0) = I_{BT0}. \end{cases}$$

Now, to minimize the objective functional:

$$J(u) = \int_0^{T_f} \eta(S_v, I_v, S_g, I_g, S_{BT}, I_{BT}, u_1, t) dt,$$
(8)

subject to the constraints

where, $\xi_i = \xi(S_v, I_v, S_g, I_g, S_{BT}, I_{BT}, u_1, t), i = 1, 2, \dots, 6$, with initial conditions:

$$S_{v}(0) = S_{v(0)}, S_{g}(0) = S_{g}(0), S_{BT}(0) = S_{BT}(0),$$

$$I_{v}(0) = I_{v(0)}, I_{g}(0) = I_{g}(0), I_{BT}(0) = I_{BT}(0).$$

The modified equation of (8) is [30]

$$\tilde{J} = \int_{0}^{T_{f}} [H_{a}(S_{v}, I_{v}, S_{g}, I_{g}, S_{BT}, I_{BT}, u_{1}, t) - \sum_{i=1}^{6} \lambda_{i} \xi_{i}(S_{v}, I_{v}, S_{g}, I_{g}, S_{BT}, I_{BT}, u_{1}, t)] dt, \qquad (9)$$

where the Hamiltonian is:

$$H_{a}(S_{v}, I_{v}, S_{g}, I_{g}, S_{BT}, I_{BT}, u_{1}, \lambda_{i}, t) = \eta(S_{v}, I_{v}, S_{g}, I_{g}, S_{BT}, I_{BT}, u_{1}, t) + \sum_{i=1}^{6} \lambda_{i} \xi_{i}(S_{v}, I_{v}, S_{g}, I_{g}, S_{BT}, I_{BT}, u_{1}, t).$$
(10)

from (9) and (10) the necessary conditions for FOCPs [35-38] are,

$$\begin{cases}
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{1} = \frac{\partial H_{a}}{\partial S_{v}}, \\
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{2} = \frac{\partial H_{a}}{\partial I_{v}}, \\
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{3} = \frac{\partial H_{a}}{\partial S_{g}}, \\
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{4} = \frac{\partial H_{a}}{\partial I_{g}}, \\
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{5} = \frac{\partial H_{a}}{\partial S_{BT}}, \\
{}^{ABC}_{0}D^{\zeta}_{t_{f}}\lambda_{6} = \frac{\partial H_{a}}{\partial I_{BT}}, \\
{}^{0}_{0} = \frac{\partial H}{\partial u},
\end{cases}$$
(11)
$$(12)$$

$$\begin{array}{l}
 ^{ABC}D_{t}^{\zeta}S_{v} = \frac{\partial H_{a}}{\partial\lambda_{1}}, \\
 ^{ABC}D_{t}^{\zeta}I_{v} = \frac{\partial H_{a}}{\partial\lambda_{2}}, \\
 ^{ABC}D_{t}^{\zeta}S_{g} = \frac{\partial H_{a}}{\partial\lambda_{3}}, \\
 ^{ABC}D_{t}^{\zeta}I_{g} = \frac{\partial H_{a}}{\partial\lambda_{4}}, \\
 ^{ABC}D_{t}^{\zeta}S_{BT} = \frac{\partial H_{a}}{\partial\lambda_{5}}, \\
 ^{ABC}D_{t}^{\zeta}I_{BT} = \frac{\partial H_{a}}{\partial\lambda_{6}}, \\
 ^{ABC}D_{t}^{\zeta}I_{BT} = \frac{\partial H_{a}}{\partial\lambda_{6}}, \\$$
 (13)

$$\lambda_j(T_f) = 0, \tag{14}$$

where $\lambda_j = 1, 2, 3, \dots, 6$ are the Lagrange Multiplies.

Theorem 4. If u_1 be the optimal controls with corresponding stats S_v^* , I_v^* , S_g^* , I_g^* , S_{BT}^* , and I_{BT}^* , then $\exists \lambda_j^*$, j = 1, 2, ..., 6, satisfies the following.

(i) Adjoint equations:

$$\begin{array}{l} {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{1} = \frac{\partial H_{a}}{\partial S_{v}} \\ = \lambda_{1} (-\alpha - \beta_{1} (1 - u_{1}) I_{BT} - \mu_{p}) + \lambda_{2} \beta_{1} (1 - u_{1}) I_{BT} \\ + \lambda_{3} \alpha - \lambda_{5} \theta_{1} u_{1} S_{BT} - \lambda_{6} \theta_{1} u_{1} I_{BT}, \\ {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{2} = \frac{\partial H_{a}}{\partial I_{v}} \\ = A_{1} - \lambda_{2} \mu_{p} - \lambda_{5} (\gamma_{1} (1 - u_{1}) S_{BT} + \theta_{1} u_{1} S_{BT}), \\ {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{3} = \frac{\partial H_{a}}{\partial S_{g}} \\ = -\lambda_{3} \beta_{2} (1 - u_{1}) I_{BT} - \lambda_{3} \mu_{p} + \lambda_{4} \beta_{2} (1 - u_{1}) I_{BT} \\ - \lambda_{5} \theta_{1} u_{1} S_{BT} - \lambda_{6} \theta_{1} u_{1} I_{BT}, \\ {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{4} = \frac{\partial H_{a}}{\partial I_{g}} \\ = A_{2} - \lambda_{4} \mu_{p} + \lambda_{5} [BN_{v} - \gamma_{2} (1 - u_{1}) S_{BT} - \theta_{1} u_{1} S_{BT}] \\ + \lambda_{6} [\gamma_{2} (1 - u_{1}) S_{BT} - \theta_{1} u_{1} I_{BT}], \\ {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{5} = \frac{\partial H_{a}}{\partial S_{BT}} \\ = \lambda_{5} [B - \gamma_{1} (1 - u_{1}) I_{v} - \gamma_{2} (1 - u_{1}) I_{g} - \theta_{1} u_{1} N_{p} - \mu_{1} \\ + \lambda_{6} [\gamma_{1} (1 - u_{1}) I_{v} + \gamma_{2} (1 - u_{1}) I_{g}], \\ {}_{0}^{ABC} D_{l_{f}}^{\zeta} \lambda_{6} = \frac{\partial H_{a}}{\partial I_{BT}} \\ = A_{3} - \lambda_{1} \beta_{1} (1 - u_{1}) S_{v} - \lambda_{2} \beta_{1} (1 - u_{1}) S_{v} \\ - \lambda_{3} \beta_{2} (1 - u_{1}) S_{g} + \lambda_{4} \beta_{2} (1 - u_{1}) S_{g} - \lambda_{6} (\theta_{1} u_{1} N_{p} - \mu_{1}). \end{array} \right\}$$

(ii) Transversality conditions:

$$\lambda_i^*(T_f) = 0, j = 1, 2, \dots, 6.$$
(16)

(iii) Optimality conditions:

$$H_a(S_v^*, I_v^*, S_g^*, I_g^*, S_{BT}^*, I_{BT}^*, u, \lambda_*) = \min_{0 \le u \le 1} H(S_v^*, I_v^*, S_g^*, I_g^*, S_{BT}^*, I_{BT}^*, u, \lambda^*).$$
(17)

*Furthermore, the control functions u*₁ *are given by,*

$$u_{1}^{*} = \max\{\min[\frac{1}{2A_{4}}(\lambda_{1}^{*}\beta_{1}S_{v}I_{BT} - \lambda_{2}^{*}\beta_{1}S_{v}I_{BT} + \lambda_{3}^{*}\beta_{2}S_{g}I_{BT} - \lambda_{4}^{*}\beta_{2}S_{g}I_{BT} + \lambda_{5}^{*}\gamma_{1}I_{v}S_{BT} + \lambda_{5}^{*}\gamma_{2}I_{g}S_{BT} - \lambda_{5}^{*}\theta_{1}S_{BT}N_{p} - \lambda_{6}^{*}\gamma_{1}I_{v}S_{BT} - \lambda_{6}^{*}\gamma_{2}I_{g}S_{BT} - \lambda_{6}^{*}\theta_{1}I_{BT}N_{p}), 1], 0\}$$
(18)

Proof. We can state that (7) using the conditions (11), where H_a^* is,

$$\begin{split} H_{a}^{*} &= A_{1} + A_{2} + A_{3} + A_{4}u_{1}^{2} + \lambda_{10}^{*ABC}D_{t}^{\zeta}S_{v}^{*} + \lambda_{20}^{*ABC}D_{t}^{\zeta}I_{v}^{*} \\ &+ \lambda_{30}^{*ABC}D_{t}^{\zeta}S_{g}^{*} + \lambda_{40}^{*ABC}D_{t}^{\zeta}I_{g}^{*} \\ &+ \lambda_{50}^{*ABC}D_{t}^{\zeta}S_{BT}^{*} + \lambda_{60}^{*ABC}D_{t}^{\zeta}I_{BT}^{*}. \end{split}$$

Moreover, $\lambda_{j}^{*}(T_{f}) = 0, j = 1, ..., 6$ holds.

The optimal control Equation (18) are proved by minimizing the condition (17). Substitute u_1^* in (7) we get,

$$\begin{split} & {}_{0}^{ABC} D^{\zeta} S_{v}^{*} = A - \alpha S_{v}^{*} - \beta_{1} (1 - u_{1}^{*}) S_{v}^{*} I_{BT}^{*} - \mu_{p} S_{v}^{*} \\ & {}_{0}^{ABC} D^{\zeta} I_{v}^{*} = \beta_{1} (1 - u_{1}^{*}) S_{v}^{*} I_{BT}^{*} - \mu_{p} I_{v}^{*} \\ & {}_{0}^{ABC} D^{\zeta} S_{g}^{*} = \alpha S_{v}^{*} - \beta_{2} (1 - u_{1}^{*}) S_{g}^{*} I_{BT}^{*} - \mu_{p} S_{g}^{*} \\ & {}_{0}^{ABC} D^{\zeta} I_{g}^{*} = \beta_{2} (1 - u_{1}^{*}) S_{g}^{*} I_{BT}^{*} - \mu_{p} I_{g}^{*} \\ & {}_{0}^{ABC} D^{\zeta} S_{BT}^{*} = B^{*} N_{v}^{*} - \gamma_{1} (1 - u_{1}^{*}) I_{v}^{*} S_{BT}^{*} - \gamma_{2} (1 - u_{1}^{*}) I_{g}^{*} S_{BT}^{*} - \theta_{1} u_{1}^{*} S_{BT}^{*} N_{p}^{*} - \mu_{1} S_{BT}^{*} \\ & {}_{0}^{ABC} D^{\zeta} I_{BT}^{*} = \gamma_{1} (1 - u_{1}^{*}) I_{v}^{*} S_{BT}^{*} - \gamma_{2} (1 - u_{1}^{*}) I_{g}^{*} S_{BT}^{*} - \theta_{1} u_{1}^{*} I_{BT}^{*} N_{p}^{*} - \mu_{1} I_{BT}^{*}. \end{split}$$

6. Numerical Results

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Here, we examine the mathematical model of *C. annuum* with AB fractional derivative and optimal control numerically. We assume initial conditions and parameter values in Table 1 with $\zeta = 0.9$ and $N(\zeta) = 1$.

The optimal control is

$$u_{1}^{*} = \max\{\min[\frac{1}{2A}(\lambda_{1}^{*}\beta_{1}S_{v}I_{BT} - \lambda_{2}^{*}\beta_{1}S_{v}I_{BT} + \lambda_{3}^{*}\beta_{2}S_{g}I_{BT} - \lambda_{4}^{*}\beta_{2}S_{g}I_{BT} + \lambda_{5}^{*}\gamma_{1}I_{v}S_{BT} + \lambda_{5}^{*}\gamma_{2}I_{g}S_{BT} - \lambda_{5}^{*}\theta_{1}S_{BT}N_{p} - \lambda_{6}^{*}\gamma_{1}I_{v}S_{BT} - \lambda_{6}^{*}\gamma_{2}I_{g}S_{BT} - \lambda_{6}^{*}\theta_{1}I_{BT}N_{p}), 1], 0\}$$

since
$$0 \le u_1 \le 1$$
. Consider $u_1 = 0.6$ and using the parametric values in (7) then,

$$\begin{split} S_{nv}(t) &= 50 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [10 - 0.07S_{(n-1)v}(t) - 0.001 \times 0.4 \times S_{(n-1)v}(t)I_{(n-1)BT}(t) \\ &- 0.03S_{(n-1)v}(t)] \\ I_v(t) &= 10 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [0.001 \times 0.4 \times S_{(n-1)v}(t)I_{(n-1)BT}(t) - 0.03I_{(n-1)v}(t)] \\ S_g(t) &= 30 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [0.07S_{(n-1)v}(t) - 0.001 \times 0.4 \times S_{(n-1)g}(t)I_{(n-1)BT}(t) \\ &- 0.03S_{(n-1)g}(t)] \\ I_g(t) &= 10 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [0.001 \times 0.4 \times S_{(n-1)g}(t)I_{(n-1)BT}(t) - 0.03I_{(n-1)g}(t)] \\ S_{BT}(t) &= 30 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [10 \times 40 - 0.025 \times 0.4 \times I_{(n-1)v}(t)S_{(n-1)BT}(t) - 0.02 \\ &\times 0.4 \times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times S_{(n-1)BT}(t) \times 80 - 0.07S_{(n-1)BT}(t)] \\ I_{BT}(t) &= 10 + \left(0.1 + \frac{(t)^{0.9}}{\Gamma(0.9)}\right) [0.025 \times 0.4 \times I_{(n-1)v}(t)S_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) + 0.02 \times 0.4 \\ &\times I_{(n-1)g}(t)S_{(n-1)BT}(t) - 0.05 \times 0.6 \times I_{(n-1)BT}(t) \times 80 - 0.07I_{(n-1)BT}(t)] \\ \end{bmatrix}$$

which gives the numerical values plotted in Figures 1–6.

In Figures 1–6, we show that the use of 60% *V. lecanni* for 5 days changed the population of susceptible and infected plants in vegetative and generative phases, as well as the variation of population of susceptible and infected white bugs.



Figure 1. Optimal control of susceptible vegetative phase $S_v(t)$ of *C. annuum*.



Figure 2. Optimal control of infected vegetative phase $I_v(t)$ of *C. annuum*.



Figure 3. Optimal control of susceptible generative phase $S_g(t)$ of *C. annuum*.



Figure 4. Optimal control of infected generative phase $I_v(g)$ of *C. annuum*.



Figure 5. Optimal control of susceptible *B. tabaci* $S_{BT}(t)$ in *C. annuum*.



Figure 6. Optimal control of infected *B. tabaci* $I_{BT}(t)$ in *C. annuum*.

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Variable with Values	Definition
$N_p = 80$	C. annuum population
$N_v = 40$	B. tabaci population
$S_v = 50$	Vegetative phase of Susceptible C. annuum
$I_v = 10$	Vegetative phase of Infected C. annuum
$S_g = 30$	Generative phase of Susceptible C. annuum
$I_g = 10$	Generative phase of Infected C. annuum
$S_{BT} = 30$	B. tabaci Susceptible insect
$I_{BT} = 10$	B. tabaci Infected insect
A = 10	Recruitment of C. annuum
B = 10	Recruitment of B. tabaci
$\alpha = 0.07$	Rate of growth from vegetative to generative phase
$\beta_1 = 0.001$	Rate of infected <i>C. annuum</i> in the vegetaive phase
$\beta_2 = 0.001$	Rate of infected <i>C. annuum</i> in the generative phase
$\gamma_1 = 0.025$	Rate of <i>B. tabaci</i> infection in the vegetaive phase
$\gamma_2 = 0.02$	Rate of <i>B. tabaci</i> infection in the generative phase
$\delta_p=0.2$	V. lecanii
$\mu_p = 0.03$	The death rate of <i>C. annuum</i>
$\mu_1 = 0.07$	Rate of natural death in <i>B. tabaci</i>
$\theta_1 = 0.05$	The death rate of <i>B. tabaci</i> due to curative intervention

Table 1. Parametric representation with approximation values of (1) in [29].

When $u_1 = 0$, i.e., without *V. lecanni*, the control system reduce to the model given below:

_

$$\begin{split} & {}^{ABC}_{0} D_{t}^{\zeta} S_{v}(t) = A - \alpha S_{v} - \beta_{1} S_{v} I_{BT} - \mu_{p} S_{v} \\ & {}^{ABC}_{0} D_{t}^{\zeta} I_{v}(t) = \beta_{1} S_{v} I_{BT} - \mu_{p} I_{v} \\ & {}^{ABC}_{0} D_{t}^{\zeta} S_{g}(t) = \alpha S_{v} - \beta_{2} S_{g} I_{BT} - \mu_{p} S_{g} \\ & {}^{ABC}_{0} D_{t}^{\zeta} I_{g}(t) = \beta_{2} S_{g} I_{BT} - \mu_{p} I_{g} \\ & {}^{ABC}_{0} D_{t}^{\zeta} S_{BT}(t) = B N_{v} - \gamma_{1} I_{v} S_{BT} - \gamma_{2} I_{g} S_{BT} - \mu_{1} S_{BT} \\ & {}^{ABC}_{0} D_{t}^{\zeta} I_{BT}(t) = \gamma_{1} I_{v} S_{BT} + \gamma_{2} I_{g} S_{BT} - \mu_{1} I_{BT}. \end{split}$$

In Figures 7–12, we compare the numerical values of the variables $S_v(t)$, $I_v(t)$, $S_g(t)$, $I_g(t)$, $S_{BT}(t)$ and $I_{BT}(t)$ with and without *V. lecanii* u_1 . The comparison of these variables of ordinary differential equations is shown in [29] of the figures are Figures 2–7. The results with *V. lecanni* in each stage of plants and insects are more effective and accurate in Atangana–Baleanu fractional derivative than ordinary differential equations stated in [29].



Figure 7. Comparison of susceptible vegetative phase $S_v(t)$ in *C. annuum* with and without *V. lecanni*.



Figure 8. Comparison of infected vegetative phase $I_v(t)$ in *C. annuum* with and without *V. lecanni*.



Figure 9. Comparison of susceptible generative phase $S_g(t)$ in *C. annuum* with and without *V. lecanni*.



Figure 10. Comparison of infected generative phase $I_v(g)$ in *C. annuum* with and without *V. lecanni*.



Figure 11. Comparison of susceptible *B. tabaci* $S_{BT}(t)$ in *C. annuum* with and without *V. lecanni*.



Figure 12. Comparison of infected *B. tabaci* $I_{BT}(t)$ in *C. annuum* with and without *V. lecanni*.

Figures 7 and 9 show that the susceptible *C. annuum* in vegetative and generative phases, which increased the population by 1% with *V. lecanni* compared to that of without *V. lecanni* because the infected *B. tabaci* cannot transmit the virus through chili plants. Figures 8 and 10 show the comparison of infected *C. annuum* in vegetative and generative phases, which decrease the population by 1% with *V. lecanni* compared to that of without *V. lecanni*.

In Figure 11, the comparison of susceptible *B. tabaci* population decreases with *V. lecanni* by 50% compared to that of without *V. lecanni* because the infected *B. tabaci* cannot infect the healthy one with an antidote. In Figure 11, the comparison of infected *B. tabaci* population decreases with *V. lecanni* by 4% compared to that of without *V. lecanni* because the infected ones were either cured or dead due to curative intervention.

In Figure 13, the measure of implementing 60% of *V. lecanni* per day will reduce 1% of infected *C. annuum* and 1% of infected *B. tabaci*. By continuing this process, the 60% of *V. lecanni* control the spread of the yellow virus within 10 days, which helps the farmers to afford the costs of cultivating the *C. annuum*.



Figure 13. Control of V. lecanni.

7. Conclusions

In this study, we described the *C. annuum* model of the yellow virus in two discrete aspects. First, we examined the *C. annuum* model and applied the optimal control. Second, we analyzed the *C. annuum* model using the Atangana–Baleanu derivative. The threshold quantity is less than one when the presented model is locally stable. Furthermore, the model is globally stable when $\psi_0 > 1$. With the help of V. lecanii (an entomopathogenic fungus), $u_1(t)$ optimal control reduced the population of infected *B. tabaci* and *C. annuum*. The numerical results of optimal conditions of the *C. annuum* model with AB-derivative are described detailly by successive approximation method. The infected population of *C. annuum* increases and decreases according to V. lecanii use and vice-versa for susceptible. The results show that using 60% of *V. lecanni* controls the spread of the yellow virus in infected *B. tabaci* and *C. annuum* over 10 days, which helps farmers to afford the costs of cultivation.

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