An efficient numerical technique for a new fractional tuberculosis model with nonsingular derivative operator

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ABSTRACT
The present study aims to investigate a new fractional model describing the dynamical behaviour of the tuberculosis infection. In this new formulation, we use a recently introduced fractional operator with Mittag–Leffler nonsingular kernel. To solve and simulate the proposed model, a new and efficient numerical method is developed based on the product-integration rule. Simulation results are provided and some discussions are given to verify the theoretical analysis. The results indicate that employing the nonsingular operator can extract the hidden aspects of the model under consideration while these features are invisible when we use the ordinary time-derivatives. Therefore, the non-integer calculus supplies more flexible models describing the asymptotic behaviour of the real-world phenomena and helps us to better understand their complex dynamics.

1. Introduction
Mathematical modelling is an important tool to describe different aspects of the natural phenomena more precisely. Among the existing excellent works in this direction, the mathematical models for different kinds of diseases have extensively attracted the attention of many scientists. One of the main advantages in this area is to identify the impressive factors preventing the spread of diseases for the purpose of treatment. This property makes this field of study more attractive for both mathematicians and biologists. Recently, the fractional calculus has been widely used to model and describe the biological processes. Some noticeable efforts have been done in [1–3] for the fractional modelling of membrane charging, nerve axon and vestibular ocular models, viscoelastic model of cells, etc. In recent years, the non-integer calculus has also been employed to investigate and discuss the spread of diseases. In [4], the malaria transmission was modelled by a fractional formalism. In [5], a time-fractional model of immunogenic tumour was studied. In [6], a fractional model of HIV infection was introduced. In [7], a fractional HIV/HCV coinfection model was employed for the effect of HIV viral load. More recent study in this direction can also be found in [8].

Tuberculosis is a bacterial disease, which is known as one of the most important causes of death in the world [9]. This disease is caused by Mycobacterium tuberculosis and spread with active tuberculosis through the air by people. A good review on the mathematical model of tuberculosis can be found in [10]. In [11], an interesting model was investigated for this infectious disease for the purpose of treatment by an optimal control strategy. In [12], a time-delay model was introduced for the tuberculosis. In [13], numerical simulations were provided using the tuberculosis data from Angola. In [14], a new tuberculosis model was developed to investigate the impact of treatment for both latently and actively infected tuberculosis populations. In [15], a tuberculosis dynamic was studied and used to reduce the infection in the Philippines’ population. As well as numerous works mentioned above, there are some studies on the fractional modelling of tuberculosis infection. In [16], a fractional multi-strain tuberculosis model was proposed and an approximation technique was developed to obtain the endemic solution. In [17], a fractional analysis was carried out to identify the transmission dynamic of a tuberculosis model. In [18], a new fractional model was introduced for the tuberculosis infection. In [19], the authors developed a fractional model to explore the impact of diabetes and resistant strains on the dynamic of tuberculosis. In [20], the tuberculosis dynamic was explored by a fractional-order model in the Caputo sense. In [21], an optimal control strategy was studied for a fractional tuberculosis model containing the effect of diabetes and resistant strains.

In recent years, the fractional calculus has gained the importance and popularity for the modelling of the
real-world problems [22–28]. However, many scientists have shown that the conventional form of the fractional derivatives with singular kernel may not be able to characterize the nonlocal dynamics in an appropriate manner; hence, to better investigate the real-world phenomena with hereditary property, there is a need to apply new types of the fractional operators with nonsingular kernel. For this purpose, a new differential operator with Mittag–Leffler (ML) kernel (ABC) has been recently developed in [29] and its main properties have been deeply investigated [30, 31]. Nowadays, this new operator has been extensively applied to describe many realistic systems [32–35]; however, the application of this operator for more practical cases (such as those in [36, 37]) should be explored and extended. In addition, more efficient numerical methods need to be improved to solve the ABC fractional models effectively. This motivates us to study the characteristics of the tuberculosis infection by using its new formulation in the fractional ABC calculus. We also develop an efficient numerical technique on the basis of the product-integration (PI) rule to solve the above-mentioned equations properly. Simulation results indicate that employing nonsingular operator can extract the hidden aspects of the model under consideration, which may be invisible when we use this model in a classic integer manner. Therefore, more flexible models are available by the non-integer calculus to investigate the real-world phenomena and understand their complex dynamics.

The rest of this paper is organized as follows. Some preliminary results are given in Section 2. Section 3 is devoted to the new fractional tuberculosis model and its main characteristics. A generalized numerical method is proposed in Section 4 solving the ABC tuberculosis model. Numerical and comparative are provided in Section 5. Finally, the paper is finished in Section 6 by some concluding remarks.

2. Preliminary results

In this section, some preliminaries are presented for the new direction of the non-integer calculus involving the nonsingular operators. Following [29], we introduce the ABC operator with ML kernel and recall its main characteristics.

Definition 2.1: Let \( 0 < t < T \) and \( \phi \in \mathbb{H}^1(0, T) \). Then, the ABC fractional order of \( \zeta \in (0, 1) \) of the function \( \phi \) is given by

\[
\begin{align*}
\frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} &= \frac{\theta}{1 - \zeta} \int_0^t E_\zeta \left[-\theta(t - \rho)^\zeta\right] \frac{d\phi(\rho)}{d\rho} \, d\rho, \\
\end{align*}
\]

(1)

where \( \theta = \zeta / (1 - \zeta) \), \( A(\zeta) \) with the property \( A(0) = A(1) = 1 \) denotes the normalization function, and \( E_\zeta \) is the one-parameter ML function. The associated integral operator is also given as

\[
\begin{align*}
\frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} &= \frac{1 - \zeta}{A(\zeta)} \phi(t) \\
&\quad + \frac{\zeta}{A(\zeta)T(\zeta)} \int_0^t (t - \rho)^{\zeta - 1} \phi(\rho) \, d\rho. \quad (2)
\end{align*}
\]

Note that, the integral operator (2) recovers the original function when \( \zeta = 0 \), while it becomes the ordinary integral for \( \zeta = 1 \).

Definition 2.2: The constant \( \phi^* \) is the equilibrium point of the system \( \frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} = f(\phi(t)) \) with the initial condition \( \phi(0) = \phi_0 \) if and only if \( f(\phi^*) = 0 \).

In the following, some useful properties are given for the ABC fractional operator [29].

Property 2.1: Let \( \phi_1, \phi_2 \in \mathbb{H}^1(0, T) \) be such that \( \frac{\mathrm{ABC}_0^\zeta \phi_1(t)}{\phi_1(t)}, \frac{\mathrm{ABC}_0^\zeta \phi_2(t)}{\phi_2(t)} \) exist a.e., and \( k_1, k_2 \in \mathbb{R} \). Then, the expression \( \frac{\mathrm{ABC}_0^\zeta (k_1 \phi_1(t) + k_2 \phi_2(t))}{\phi_1(t)} \) exists a.e., and

\[
\begin{align*}
\frac{\mathrm{ABC}_0^\zeta (k_1 \phi_1(t) + k_2 \phi_2(t))}{\phi_1(t)} &= k_1 \frac{\mathrm{ABC}_0^\zeta \phi_1(t)}{\phi_1(t)} + k_2 \frac{\mathrm{ABC}_0^\zeta \phi_2(t)}{\phi_2(t)}. \quad (3)
\end{align*}
\]

Property 2.2: For a constant function \( \phi(t) \equiv c \), the ABC operator is zero, i.e.

\[
\frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} = 0. \quad (4)
\]

Property 2.3: The Laplace transform of the ABC differential operator is in the form

\[
\mathcal{L} \left[ \frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} \right] = \frac{A(\zeta)}{1 - \zeta} \frac{p^2 \Sigma [\phi(t)](p)}{p^2 + \frac{\zeta}{1 - \zeta}}. \quad (5)
\]

Property 2.4: The following useful relation is satisfied between the operators (1) and (2)

\[
\frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} = \phi(t) - \phi(0). \quad (6)
\]

Property 2.5: For each \( \phi \in \mathbb{H}^1(0, T) \), the operator (1) satisfies

\[
\left\| \frac{\mathrm{ABC}_0^\zeta \phi(t)}{\phi(t)} \right\| \leq \frac{A(\zeta)}{1 - \zeta} \|\phi(t)\|, \quad (7)
\]

where \( \|\phi(t)\| = \max_{0 \leq t \leq T} |\phi(t)| \).

Property 2.6: The ABC operator (1) fulfills the Lipschitz condition, i.e. for each \( \phi_1, \phi_2 \in \mathbb{H}^1(0, T) \) there exists a constant \( L > 0 \) such that

\[
\left\| \frac{\mathrm{ABC}_0^\zeta \phi_1(t)}{\phi_1(t)} - \frac{\mathrm{ABC}_0^\zeta \phi_2(t)}{\phi_2(t)} \right\| \leq L \|\phi_1(t) - \phi_2(t)\|. \quad (8)
\]

Some additional information is available in [29, 38], which the interested readers can refer to.
3. The new fractional tuberculosis model

This section develops a new fractional version of a tuberculosis model in the ABC sense with incomplete treatment. The original version of this model including integer-order derivatives has been previously examined in [39]; however, the influence of hereditary, as a fundamental feature of many biological processes, has not been included in this model. To overcome this drawback, we employ the newly introduced ABC operator (1) instead of ordinary time-derivative in the tuberculosis model since the ABC operator essentially includes the effect of memory in its definition. In addition, according to [40], we modify the fractional operator by an auxiliary parameter \( \sigma \), having the dimension of sec., to ensure that the right- and left-hand sides of the resultant equation possess the same dimension \( s^{-1} \). Considering these features, the new transmission model of the tuberculosis infection takes the form

\[
\begin{align*}
\frac{1}{\sigma + 1} ABC_0^\sigma D_0^\tau S(t) &= \Lambda - \mu S(t) - bS(t)l(t), \quad t \geq 0, \\
\frac{1}{\sigma + 1} ABC_0^\sigma D_0^\tau L(t) &= -((\mu + \varepsilon)l(t) + (1 - k)\delta T(t) + b(l(t)S(t), \quad t \geq 0, \\
\frac{1}{\sigma + 1} ABC_0^\sigma D_0^\tau I(t) &= -(\mu + \gamma + \alpha_1)l(t) + k\delta T(t) + \varepsilon l(t), \quad t \geq 0, \\
\frac{1}{\sigma + 1} ABC_0^\sigma D_0^\tau T(t) &= -(\mu + \delta + \alpha_2)T(t) + \gamma l(t), \quad t \geq 0,
\end{align*}
\]

along with the initial conditions

\[
S(0) = S_0, \quad L(0) = L_0, \quad l(0) = l_0, \quad T(0) = T_0. \quad (10)
\]

As can be seen, the fractional operator in Equation (9) has been taken in the sense of ABC with \( \zeta \in (0, 1) \). A physical description of the parameters and variables is given as below. The total population is partitioned into four distinct categories depending on their epidemiological stages. The first category, which is denoted by \( S(t) \), includes the susceptible individuals. This population is never infected by the tuberculosis. The second compartment is denoted by \( L(t) \) and contains the latent infected persons. Although these people have been infected, they do not have the symptoms of the disease, and they are not infectious as well. The next compartment is the population of infectious individuals, which is signified by \( l(t) \), has active tuberculosis, can transmit the infection, and is not in treatment. The last group, denoted by \( R(t) \), is the population of infected individuals, which is under treatment. Since the state variables in (9) reflect the human population, they are all assumed to be non-negative. From the first equation in (9), the recruitment of individuals can increase the susceptible population at a rate \( \Lambda \). The coefficient \( \mu \) is the natural death rate of all individuals. The contact between susceptible and infectious individuals acquire the tuberculosis infection within a transmission coefficient \( b \).

After being infected, the individual first enters the latent infected population at a rate \( b(l(t)S(t), and then becomes infectious at a rate \( \varepsilon \). By detecting an infectious individual, he/she will be treated and enter the treatment compartment. The parameter \( \gamma \) is the per-capita treatment rate of infected population. At a rate \( \delta \), a treated individual leaves his/her population. However, the treatment may fail and the treated person may enter the compartment \( I \); otherwise, due to the remainder of *Mycobacterium tuberculosis*, the treated individual becomes latently infectious and enters the population \( L \).

The failure of the treatment is reflected by the parameter \( k (0 \leq k \leq 1) \), where \( k = 1 \) denotes the complete failure of the treatment, i.e. all the treated individuals are still infectious, while \( k = 0 \) expresses that all the treated persons have become latently infectious. The parameters \( \alpha_1 \) and \( \alpha_2 \) also denote the tuberculosis-induced death rate of the infectious and under treatment individuals, respectively.

\[\text{3.1. Non-negative solution}\]

In this section, we show that the feasibility region of the system (9) is given by the closed set

\[
\Omega = \left\{ (S, L, I, T) \in \mathbb{R}_+^4 : S + L + I + T \leq \frac{\Lambda}{\mu} \right\}, \quad (11)
\]

which is positively invariant. For this purpose, we consider and prove the following lemma.

**Lemma 3.1:** The closed set \( \Omega \) in (11) is positively invariant with respect to the fractional model (9).

**Proof:** By adding all relations in (9), we obtain the fractional derivative of total population as follows

\[
\frac{1}{\sigma + 1} ABC_0^\sigma D_0^\tau P(t) = \Lambda - \mu P(t) - a_1l(t) - a_2T(t) \leq \Lambda - \mu P(t), \quad (12)
\]

where \( P(t) = S(t) + L(t) + I(t) + T(t) \). Applying the Laplace transform, Equation (12) provides

\[
P(t) \leq \left( \frac{A(\zeta)}{A(\zeta) + (1 - \zeta)\mu} P(0) + \frac{(1 - \zeta)\Lambda}{A(\zeta) + (1 - \zeta)\mu} E_{\zeta,1}(-mt^\zeta) \right) + \frac{\zeta \Lambda}{A(\zeta) + (1 - \zeta)\mu} E_{\zeta,\zeta+1}(-mt^\zeta), \quad (13)
\]

where \( m = \zeta \mu / (A(\zeta) + (1 - \zeta)\mu), \quad \Lambda = \sigma^{1 - \zeta}, \quad \mu = \sigma^{1 - \zeta} \mu, \quad \text{and} \ E_{ab} \) denotes the two-parameter ML function. Based on the fact that the ML function has an asymptotic behaviour, we obtain \( P(t) \leq \Lambda / \mu \) as \( t \) tends to infinity. Thus, the entire solution of the
fractional model (9) remains in Ω for every t > 0 when the initial condition belongs to Ω. Consequently, the closed set Ω is a positively invariant set according to the fractional model (9).

### 3.2. Equilibrium points and basic reproduction number

The equilibrium points of the system (9) are obtained by setting

\[
\frac{1}{\sigma^{1-\zeta}}ABC_0D_z^\zeta S(t) = \frac{1}{\sigma^{1-\zeta}}ABC_0D_z^\zeta L(t) = \frac{1}{\sigma^{1-\zeta}}ABC_0D_z^\zeta I(t) = 0.
\]

This equation implies that the system (9) always has a disease-free equilibrium point \( E_0 = (\Lambda/\mu, 0, 0, 0) \). In addition, when the endemic threshold \( R_0 \) is greater than 1,

\[
R_0 = \frac{(b \varepsilon \Delta \mu)(\mu + \delta + a_2)}{(\mu + \varepsilon)(\mu + \gamma + a_1)(\mu + \delta + a_2)} - (\gamma \delta)(k(\mu + \varepsilon) + (1 - k)\varepsilon)
\]

the system (9) possesses another endemic steady-state \( E_e = (S_e, L_e, I_e, T_e) \),

\[
S_e = \frac{\Lambda}{\mu + b l_e}, \\
L_e = \frac{1}{\varepsilon}(\mu + \gamma + a_1) - \frac{\gamma \delta}{\mu + \delta + a_2} l_e, \\
I_e = \frac{\mu}{b}(R_0 - 1), \\
T_e = \frac{\gamma}{\mu + \delta + a_2} l_e.
\]

Note that, the endemic threshold \( R_0 \) in Equation (15), which is also known as the basic reproduction number, is obtained by the next generation matrix technique described in [41]. To do so, let us consider \( Z(t) = (L(t), I(t), T(t), S(t)) \) and write the system (9) in a compact form

\[
ABC_0D_z^\zeta Z(t) = F(Z(t)) - V(Z(t)),
\]

where

\[
F(Z(t)) = \sigma^{1-\zeta} \begin{bmatrix} b(t)S(t) \\ 0 \\ 0 \\ 0 \end{bmatrix},
\]

\[
V(Z(t)) = \sigma^{1-\zeta} \begin{bmatrix} (\mu + \varepsilon)\mu T(t) - (1 - k)\delta T(t) \\ (\mu + \gamma + a_1)\mu I(t) - k\delta T(t) - \varepsilon L(t) \\ (\mu + \delta + a_2)T(t) - \gamma I(t) \\ -\Lambda + \mu S(t) + bS(t)I(t) \end{bmatrix}.
\]

For \( F(Z(t)) \) and \( V(Z(t)) \), the Jacobian matrices at \( E_0 \) are, respectively, obtained as

\[
J_f(E_0) = \sigma^{1-\zeta} \begin{bmatrix} M_{3 \times 3} & 0 \\ 0 & 0 \end{bmatrix},
\]

\[
J_v(E_0) = \sigma^{1-\zeta} \begin{bmatrix} N_{3 \times 3} & 0 \\ 0 & \frac{\mu}{b} & 0 \end{bmatrix},
\]

where

\[
M = \begin{bmatrix} 0 & \frac{b \Delta}{\mu} & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix},
\]

\[
N = \begin{bmatrix} \mu + \varepsilon & 0 & -(1 - k)\delta \\ -\varepsilon & \mu + \gamma + a_1 & -k\delta \\ 0 & -\gamma & \mu + \delta + a_2 \end{bmatrix}.
\]

For the system (9), the next generation matrix is \( FV^{-1} \), and thus, the basic reproduction number (15) is obtained from \( R_0 = \rho(FV^{-1}) \). From the epidemiological viewpoint, this parameter \( R_0 \) is interpreted as the expected number of secondary cases produced by a typical infectious individual in a completely susceptible population.

### 4. Numerical method

For the fractional tuberculosis model (9), the exact solution may not be available in general; hence, a new numerical method based on the product-integration (PI) rule [42] is developed to provide a precise approximate solution. For this purpose, first we consider the initial-value problem

\[
ABC_0D_z^\zeta \phi(t) = f(\phi(t), t), \\
\phi(0) = \phi_0,
\]

where \( f(\phi(t), t) \) is a continuous function. By applying the integral operator (2) to the both sides of Equation (21) and using the Property 2.4 in Equation (6), the following Volterra integral equation is derived

\[
\phi(t) - \phi(0) = 1 - \frac{\varepsilon}{A(\zeta)}f(\phi(t), t) + \frac{\varepsilon}{A(\zeta)\Gamma(\zeta)} \int_{0}^{t} (t - \rho)^{\zeta-1} f(\phi(\rho), \rho) \, d\rho.
\]

By putting \( t = n\varepsilon \) in Equation (22), in which \( \varepsilon \) is the time step size, we obtain

\[
\phi(t_n) \approx \phi(0) + 1 - \frac{\varepsilon}{A(\zeta)}f(\phi(t_n), t_n) + \frac{\varepsilon}{A(\zeta)\Gamma(\zeta)} \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} (t_n - \rho)^{\zeta-1} f(\phi(\rho), \rho) \, d\rho.
\]
Now, the function \( f(\phi(\rho), \rho) \) is approximated by the first-order Lagrange interpolation

\[
f(\phi(\rho), \rho) \approx f(\phi(t_{i+1}), t_{i+1}) + \frac{\rho - t_{i+1}}{h} \times (f(\phi(t_{i+1}), t_{i+1}) - f(\phi(t_i), t_i)), \quad \rho \in [t_i, t_{i+1}],
\]

where the notation \( \phi_t \) denotes the quantity \( \phi(t_i) \). Substituting (24) into (23) and doing some algebraic manipulations, the PI formula in the ABC sense (ABC-PI) is derived in the form below

\[
\phi_n = \phi_0 + \frac{\zeta h^\zeta}{A(\zeta)} \left( \alpha_n f(\phi_0, t_0) + \sum_{j=1}^{n} \beta_{n-j} f(\phi_j, t_j) \right), \quad n \geq 1,
\]

where

\[
\alpha_n = \frac{(n-1)\zeta^{n-1} - n^\zeta (n - \zeta - 1)}{\Gamma(\zeta + 2)},
\]

\[
\beta_j = \begin{cases} 1 \frac{1 - \zeta}{\Gamma(\zeta + 2)}, & j = 0, \\ \frac{(j - 1)\zeta^{j-1} - j \zeta^{j+1}}{\Gamma(\zeta + 2)}, & j = 1, 2, \ldots, n - 1. \end{cases}
\]

Note that, according to the analysis in [43], the error here satisfies \( |\phi(t_n) - \phi_n| = O(h^{1+\zeta}) \), i.e. the order of convergence is \( 1 + \zeta \). Applying of the proposed scheme to the system (9), we attain the recursive formulas

\[
\phi_{1,n} = \phi_{1,0} + \frac{\zeta h^\zeta}{A(\zeta)} \left( \alpha_n f_1(\phi_{1,0}, \phi_{2,0}, \phi_{3,0}, \phi_{4,0}) + \sum_{j=0}^{n} \beta_{n-j} f_1(\phi_{1,j}, \phi_{2,j}, \phi_{3,j}, \phi_{4,j}) \right),
\]

\[
\phi_{2,n} = \phi_{2,0} + \frac{\zeta h^\zeta}{A(\zeta)} \left( \alpha_n f_2(\phi_{1,0}, \phi_{2,0}, \phi_{3,0}, \phi_{4,0}) + \sum_{j=0}^{n} \beta_{n-j} f_2(\phi_{1,j}, \phi_{2,j}, \phi_{3,j}, \phi_{4,j}) \right),
\]

\[
\phi_{3,n} = \phi_{3,0} + \frac{\zeta h^\zeta}{A(\zeta)} \left( \alpha_n f_3(\phi_{1,0}, \phi_{2,0}, \phi_{3,0}, \phi_{4,0}) + \sum_{j=0}^{n} \beta_{n-j} f_3(\phi_{1,j}, \phi_{2,j}, \phi_{3,j}, \phi_{4,j}) \right),
\]

\[
\phi_{4,n} = \phi_{4,0} + \frac{\zeta h^\zeta}{A(\zeta)} \left( \alpha_n f_4(\phi_{1,0}, \phi_{2,0}, \phi_{3,0}, \phi_{4,0}) + \sum_{j=0}^{n} \beta_{n-j} f_4(\phi_{1,j}, \phi_{2,j}, \phi_{3,j}, \phi_{4,j}) \right),
\]

where \( \phi_1, \phi_2, \phi_3, \phi_4 \) are correspondent to the variables \( S, L, I, T \), respectively, and

\[
f_1(\phi_1, \phi_2, \phi_3, \phi_4) = \Lambda - \mu_1 \phi_1 - b \phi_1 \phi_3,
\]

\[
f_2(\phi_1, \phi_2, \phi_3, \phi_4) = - (\mu + \varepsilon) \phi_2 + (1 - k) \delta \phi_4 + b \phi_3 \phi_1,
\]

\[
f_3(\phi_1, \phi_2, \phi_3, \phi_4) = - (\mu + r + a_1) \phi_3 + k \delta \phi_4 + \varepsilon \phi_2,
\]

\[
f_4(\phi_1, \phi_2, \phi_3, \phi_4) = - (\mu + \delta + a_2) \phi_4 + \gamma \phi_3.
\]

5. Results and discussion

In this section, the proposed scheme in Section 4 is used to numerically simulate the fractional-order tuberculosis model (9). The dynamical behaviour of the new model is also studied for different non-integer order \( \zeta \). The results in the ABC sense are compared with those of a classic integer model [39] and a fractional model with conventional Caputo derivative [17]. The values of biological parameters used in these simulations are selected according to a realistic analysis in different literature [39, 44–48]. Table 1 reports these values including appropriate citations. Without loss of generality, we take the order of fractional operator and its modification parameter as \( \zeta = 0.93, 0.95, 0.97, 0.99 \) and \( \sigma = 0.99 \), respectively. The initial conditions are also selected as \((S(0), L(0), I(0), T(0)) = (0.8, 0.05, 0.1, 0.05)\) [17].

Figures 1–4 display the approximate solutions within two different fractional operators and various values of \( \zeta \) as compared to the classic integer solution. Note that, in this case, the basic reproduction number and the disease-free equilibrium are \( R_0 = 0.7121 < 1 \) and \( E_0 = (55445, 0, 0, 0) \), respectively. As can be seen, the state trajectories converge to the disease-free steady-state for all cases as \( t \to \infty \). In addition, the solution of the fractional models tends to the classic integer solution when \( \zeta \to 1 \). In the next simulation, we change the parameter \( b = 5 \times 10^{-3} \) in order to get a bigger value than 1 for \( R_0 \) to reach an endemic state, while the other parameters remain unchanged as in Table 1.

In this case, we have the following values for \( R_0 \) and the endemic steady-state \( R_0 = 7.1206 > 1 \) and \( E_0 = (7787, 43458, 175, 78) \), respectively. Simulation results for this new case are depicted in Figures 5–8, which show the convergence of the fractional-order models to the endemic equilibrium. However, the difference

| Table 1. The realistic values of parameters. |
|--------------------------------------------|
| Parameter   | Value     |
| \( \Lambda \) | 792.8571 [39] |
| \( b \)      | \( 5 \times 10^{-3} \) [44] |
| \( \mu \)    | \( \frac{1}{10} \) [45] |
| \( k \)      | 0.15 [39] |
| \( \delta \) | 1.5 [46] |
| \( \varepsilon \) | 0.00368 [47] |
| \( \gamma \) | 0.7 [39] |
| \( a_1 \)    | 0.3 [48] |
| \( a_2 \)    | 0.05 [39] |
Figure 1. The stability of the disease-free steady-state ($\zeta = 0.93$ and $R_0 < 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) under treatment individuals.

Figure 2. The stability of the disease-free steady-state ($\zeta = 0.95$ and $R_0 < 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) under treatment individuals.
Figure 3. The stability of the disease-free steady-state ($\zeta = 0.97$ and $R_0 < 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) under treatment individuals.

Figure 4. The stability of the disease-free steady-state ($\zeta = 0.99$ and $R_0 < 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) under treatment individuals.
Figure 5. The stability of the endemic stead-state ($\zeta = 0.93$ and $R_0 > 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) under treatment individuals.

Figure 6. The stability of the endemic stead-state ($\zeta = 0.95$ and $R_0 > 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) Under treatment individuals.
Figure 7. The stability of the endemic steady-state ($\zeta = 0.97$ and $R_0 > 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) Under treatment individuals.

Figure 8. The stability of the endemic steady-state ($\zeta = 0.99$ and $R_0 > 1$). (a) Susceptible individuals. (b) Latently infected individuals. (c) Actively infectious individuals and (d) Under treatment individuals.
between the objective ABC and Caputo asymptotic behaviours is obvious by the figures, and both the fractional solutions converge to the classic integer solution as $\xi \to 1$. Therefore, the fractional operator itself can be considered a degree of freedom in the modelling procedure to extract hidden aspects of the real-world dynamics. This property, i.e. flexibility of the model, is one of the main advantages of the fractional calculus over the classic integer-order models. Indeed, compared to the power function kernel of the classical Caputo derivative, the ML type kernel in the ABC definition provides a remarkable chance to analyse various natural and physical phenomena in the real-world systems.

6. Conclusions
The main objective of this research was to investigate a new fractional tuberculosis model involving a non-singular operator with ML kernel. The proposed model was solved by a new and efficient numerical algorithm based on the PI rule. Numerical results in Figures 1–8 verified the usefulness of employing nonsingular operator to extract different aspects of the considered model compared to the other integer and non-integer order operators. Therefore, the fractional calculus has the potential of providing more flexible models than the classic integer calculus to describe the complex dynamics of the real-world phenomena.

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