ANWAR ZEB, GUL ZAMAN, M. IKHLAQ CHOHAN, SHAHER MOMANI, VEDAT SUAT ERTÜRK

Abstract. In this paper, we consider the SIRC (Susceptible-Infected-Recovered-Cross immune) epidemic model. First the non-negative solution of the SIRC model in fractional order is presented. Then the multi-step generalized differential transform method (MSGDTM) is employed to compute an approximation to the solution of the model of fractional order. The obtained results are compared with the results by fourth order Runge-Kutta method and nonstandard numerical method in the integer form. Finally, we present some numerical results.

1. Introduction

Infectious agents have had decisive influence on the history of mankind. In fourteenth century Black Death has taken lives of about one third of Europe’s population [1]. The first major epidemic in the USA was Yellow Fever epidemic in Philadelphia in 1793, in which about 5,000 human died out of total 50,000 population. Epidemic models are used to understand the spread of infectious diseases in population [1, 2]. The practical use of epidemic models must rely heavily on the realism put into the models. This does not mean that a reasonable model can include all possible effects but rather incorporate the mechanisms in the simplest possible fashion so as to maintain major components that influence disease propagation. Great care should be taken before epidemic models are used for prediction of real phenomena [3]. However, even simple models should, and often pose important questions about the underlying mechanisms of infection spread and possible means of control of the disease or epidemic. Kermack and McKendrick first time introduced an epidemic model see for more detail [2]. These papers have had a major influence on the development of mathematical models to capture the spread of different diseases and are still useful in many epidemic situations. The models presented in these papers laid out a foundation for modeling infectious diseases in constant population. It means that there is no birth or death from infection. Kermack and McKendrick [2] in their first paper start with the assumption that all members of the community are initially equally susceptible to the disease, and that a complete immunity is conferred after the infection. The population is divided into three distinct classes: $S$ represents healthy individuals who can catch the disease; $I$ represents the number of infected individuals; $R$ represents the recovered

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(removed) individuals who have had the disease and are now immune to the infection (or removed from further propagation of the disease by some other means). These subdivisions of the population are called compartments. To formulate in regular order, the individual goes through consecutive states like $S \rightarrow I \rightarrow R$, such models are often called the SIR models.

Two epidemic models: the susceptible-infected-susceptible (SIS) model and the susceptible infected-recovered (SIR) model are commonly used for studying the spreads of epidemics \[3,4\]. In the SIS epidemic model a recovered individual can be infected again while in the SIR model which assume recovered individuals have lifelong immunity to the disease and this difference makes them suitable for different kinds of infectious diseases. For instance, childhood diseases in which individuals can have long-lasting immunity, either naturally or from vaccination, are appropriate for SIR model. While for viruses transmitting infection, it is more reasonable to use SIS model. Several researchers considered both SIS and SIR epidemic models and presented different epidemics in different regions all over the world \[3-5\]. Hethcote \[6\], presented the interaction of susceptible $S(t)$, infected $I(t)$ and recovered $R(t)$ individuals is given by:

\[
\frac{dS(t)}{dt} = N(t) - \mu S(t) - \beta S(t)I(t),
\]

\[
\frac{dI(t)}{dt} = \beta S(t)I(t) - (\gamma + \mu)I(t),
\]

\[
\frac{dR(t)}{dt} = \gamma I(t) - \mu R(t).
\]

Here $N(t)$ is the total population, $\beta$ is the interaction rate of infection, $\mu$ is the death rate and $\gamma$ is the recovery rate. Zaleta and Henandez \[7\] considered a simple two dimensional SIS model with vaccination showing backward bifurcation. Casagrandi et al. \[9\] presented the SIRC (Susceptible-Infected-Recovered-Cross-immune) model. This compartment (C) presents an intermediate state between the fully susceptible (S) and the fully protected (R) one. For numerical solutions Jodar et al. \[9\] presented the nonstandard finite difference schemes. Also Samanta \[10\] extended the work of Casagrandi et al. \[9\] for time dependent population size and distributed time delay. But all these work has been done in the integer order differential equations. Nowadays several researchers work on the fractional order differential equations because of best presentation of many phenomena. Fractional calculus represents a generalization of the ordinary differentiation and integration to non-integer and complex order. Fractional calculus is used to established new models in many fields not only in mathematics see for
example [11-16]. For this purpose Shahed and Alseadi [17] developed a fractional SIRC model. In their work they presented a detailed analysis for the asymptotic stability of disease-free and positive fixed point.

In this paper, we consider an SIRC model. First we show the positive solution of SIRC model in fractional order. Then we use the multi-step generalized differential transform method to approximate the numerical solution. Finally we compare our numerical results with nonstandard numerical method and forth order Runge-Kutta method.

This paper is organized as: In Section 2, we present formulation of the model with some basic definitions and notations related to this work. In Section 3, we show the non-negative solution and uniqueness of the model. In Section 4, the multi-step generalized differential transform method (MSGDTM) is applied to the model. In Section 5, the numerical simulations are presented graphically. Finally, we give conclusion.

2. Formulation of Model with Preliminaries

Here, we consider the model taking by M. El-Shahed et al. [17].

\[
\begin{align*}
\frac{dS(t)}{dt} &= \mu (1 - S(t)) - \beta I(t)S(t) + \gamma C(t), \\
\frac{dI(t)}{dt} &= \beta I(t)S(t) + \sigma \beta C(t)I(t) - (\mu + \theta)I(t), \\
\frac{dR(t)}{dt} &= (1 - \sigma) \beta C(t)I(t) + \theta I(t) - (\mu + \delta)R(t), \\
\frac{dC(t)}{dt} &= \delta R(t) - \beta C(t)I(t) - (\mu + \gamma)C(t).
\end{align*}
\]

With \( S(0) = S_0, \ I(0) = I_0, \ R(0) = R_0, \ C(0) = C_0 \).

Here \( \beta \) is the contact rate of infection, \( \gamma^{-1} \) is the cross-immune period, \( \theta^{-1} \) is the infectious period, \( \delta^{-1} \) is the total immune period and \( \sigma \) is the fraction of the exposed cross-immune individuals who are recruited in a unit time into the infective subpopulation. The total population \( N(t) = S(t) + I(t) + R(t) + C(t) \).

so we obtain by adding all equations of the system (1)
\[
\frac{d N(t)}{d t} = \mu (1 - N(t)).
\] (2)

Now we introduced fractional order to the system (1) which is consisting of ordinary differential equations. The new system is described by the following set of fractional order differential equations:

\[
D_t^\alpha S(t) = \mu (1 - S(t)) - \beta S(t) I(t) + \gamma C(t),
\] (3)

\[
D_t^\alpha I(t) = \beta S(t) I(t) + \sigma \beta C(t) I(t) - (\mu + \theta) I(t),
\] (4)

\[
D_t^\alpha R(t) = (1 - \sigma) \beta C(t) I(t) + \theta I(t) - (\mu + \delta) R(t),
\] (5)

\[
D_t^\alpha C(t) = \delta R(t) - \beta C(t) I(t) - (\mu + \gamma) C(t),
\] (6)

\[
D_t^\alpha N(t) = \mu - \mu N(t).
\] (7)

Here we used the Caputo sense fractional derivative \(D_t^\alpha\).

Now we give some basic definitions related to this work and fractional calculus [11-16].

**Definition** A function \(f(x) (x > 0)\) is said to be in the space \(C_\alpha (\alpha \in \mathbb{R})\) if it can be written as for some \(p > \alpha\) where \(f_1(x)\) is continuous in \([0, \infty)\), and it is said to be in space \(\mathbb{C}_\alpha^\alpha\) if \(f^{(\alpha)} \in C_\alpha, m \in \mathbb{N}\).

**Definition** The Riemann-Liouville integral operator of order \(\alpha > 0\) with \(a \geq 0\) is defined as

\[
(J_\alpha f)(x) = \frac{1}{\Gamma(\alpha)} \int_a^x (x - t)^{\alpha-1} f(t) \, dt, \quad x > a,
\] (8)

\[
(J_0^\alpha f)(x) = f(x).
\] (9)

Properties of the above operator can be found in [11]. We only need the following:

**Definition** For \(f \in C_\alpha\), and for \(\alpha, \beta > 0, a \geq 0, c \in \mathbb{R}\) and \(\gamma > -1\), we have

\[
(J_\alpha J_\alpha f)(x) = (J_\alpha J_\alpha f)(x) = (J_\alpha^{\alpha+\beta} f)(x),
\] (10)

\[
J_\alpha^\alpha x^\gamma = \frac{x^{\gamma+\alpha}}{\Gamma(\alpha)} B \left( \frac{x-a}{x}, \alpha, \gamma + 1 \right),
\] (11)
where \( B_{\tau}(\alpha, \gamma + 1) \) is incomplete beta function which is defined as

\[
B_{\tau}(\alpha, \gamma + 1) = \int_{0}^{\tau} t^{\alpha-1}(1-t)^{\gamma} \, dt,
\]

(12)

\[
J_a^\alpha e^{cx} = e^{a c} (x-a)^\alpha \sum_{k=0}^{\infty} \frac{(c(x-a))^k}{\Gamma(\alpha+k+1)}.
\]

(13)

**Definition** The Caputo fractional derivative of \( f(x) \) of order \( \alpha > 0 \) with \( a \geq 0 \) is defined as

\[
(D_a^\alpha f)(x)(J_a^{m-a} f^{(m)})(x) = \frac{1}{\Gamma(m-\alpha)} \int_{0}^{x} (x-t)^{\alpha+1-m} f^{(m)}(t) \, dt,
\]

(14)

for \( m-1 < \alpha \leq m, m \in \mathbb{N}, x \geq a, f(x) \in C^n_a \).

The fractional derivative was investigated by many authors, for \( m-1 < \alpha \leq m, f(x) \in C^n_a \) and \( \alpha \geq -1 \), we have

\[
(J_a^\alpha D_a^\alpha f)(x) = J_a^{m} D_a^{m} f(x) = f(x) - \sum_{k=0}^{m-1} f^{(k)}(a) \frac{(x-a)^k}{k!}.
\]

(15)

For mathematical properties of fractional derivatives and integrals one can consult the mentioned references.

3. Non-negative solutions

Let \( X^s = \{ X \in R^3 : X \geq 0 \} \) and \( X(t) = (S(t), I(t), R(t), C(t), N(t))^T \). For the proof of the theorem about non-negative solutions we shall need the following Lemma [16]:

**Lemma** (Generalized Mean Value Theorem) Let \( f(x) \in C[a, b] \) and \( D^\alpha f(x) \in C[a, b] \) for \( 0 < \alpha \leq 1 \).

Then we have,

\[
f(x) = f(a) + \frac{1}{\Gamma(\alpha)} D_a^\alpha f(\xi)(x - a)^\alpha
\]

(16)

with \( 0 \leq \xi \leq x \), for all \( x \in (a, b] \).

**Remark** Suppose \( f(x) \in C[a, b] \) and \( D^\alpha f(x) \in C[a, b] \) for \( 0 < \alpha \leq 1 \). It is clear from the
Above Lemma that if \( D^\alpha f(x) \geq 0 \), for all \( x \in (0, b) \), then the function \( f \) is non-decreasing, and if \( D^\alpha f(x) \leq 0 \), for all \( x \in (0, b) \), then the function \( f \) is non-increasing.

**Theorem** There is a unique solution for the initial value problem given by (3)-(7), and the solution remains in \( \mathbb{R}^+ \).

**Proof** The existence and uniqueness of the solution of (3)-(7), in \((0, \infty)\) can be obtained from [16, Theorem 3.1 and Remark 3.2]. We need to show that the domain \( \mathbb{R}^+ \) is positively invariant. Since

\[
D^\alpha S \big|_{x=0} = \mu + \gamma C \geq 0, \\
D^\alpha I \big|_{x=0} = 0, \\
D^\alpha R \big|_{x=0} = (1 - \sigma)\beta C I + \theta I \geq 0, \\
D^\alpha C \big|_{x=0} = \delta R \geq 0, \\
D^\alpha N \big|_{x=0} = \mu \geq 0.
\]

On each hyper-plane bounding the nonnegative orthant, the vector field points into \( \mathbb{R}^+ \).


We applying the multi-step generalized differential transform method to fined the approximate solution of equations (3)-(7), which gives an accurate solution over a longer time frame as compared to the standard generalized differential transform method. Taking the differential transform of equations (3)-(7) with respect to time we obtain,
ANALYTIC NUMERIC SOLUTION FOR SIRC EPIDEMIC MODEL IN FRACTIONAL ORDER

\[
S(k + 1) = \frac{\Gamma(\alpha k + 1)}{\Gamma((\alpha k + 1) + 1)} \left( \mu (1 - S(k)) - \beta \sum_{i=0}^{k} S(k-i) I(i) + \gamma C(k) \right),
\]

\[
I(k + 1) = \frac{\Gamma(\alpha k + 1)}{\Gamma((\alpha k + 1) + 1)} \left( \beta \sum_{i=0}^{k} S(k-i) I(i) + \sigma \sum_{j=0}^{k} C(k-j) I(j) - (\mu + \theta) I(k) \right),
\]

\[
R(k + 1) = \frac{\Gamma(\alpha k + 1)}{\Gamma((\alpha k + 1) + 1)} \left( (1 - \sigma) \beta \sum_{i=0}^{k} C(k-i) I(i) + \theta I(k) - (\mu + \delta) R(k) \right),
\]

\[
C(k + 1) = \frac{\Gamma(\alpha k + 1)}{\Gamma((\alpha k + 1) + 1)} \left( \delta R(k) - \beta \sum_{i=0}^{k} C(k-i) I(i) - (\mu + \gamma) C(k) \right),
\]

\[
N(k + 1) = \frac{\Gamma(\alpha k + 1)}{\Gamma((\alpha k + 1) + 1)} \left( \mu - \mu N(k) \right).
\]

Here \( S(k), I(k), R(k), C(k) \) and \( N(k) \) are the differential transformation of \( S(t), I(t), R(t), C(t) \) and \( N(t) \). The differential transform of the initial conditions are \( S(0) = S_0, I(0) = I_0, R(0) = R_0, C(0) = C_0 \) and \( N(0) = N_0 \).

In view of the differential inverse transform, the differential transform series solution for the system can be obtained as

\[
\begin{align*}
S(t) &= \sum_{k=0}^{K} S(k) t^{\alpha k} \\
I(t) &= \sum_{k=0}^{K} I(k) t^{\alpha k} \\
R(t) &= \sum_{k=0}^{K} R(k) t^{\alpha k} \\
C(t) &= \sum_{k=0}^{K} C(k) t^{\alpha k} \\
N(t) &= \sum_{k=0}^{K} N(k) t^{\alpha k}.
\end{align*}
\]

Now according to the multi-step generalized differential transform method the series solution for the equations (3)-(7) is suggested by
\[
S(t) = \begin{cases} 
\sum_{k=0}^{K} S_1(k) t^a k, & t \in [0, t_1] \\
\sum_{k=0}^{K} S_2(k) (t - t_1)^a k, & t \in [t_1, t_2] \\
\sum_{k=0}^{K} S_M(k) (t - t_{M-1})^a k, & t \in [t_{M-1}, t_M] 
\end{cases}
\]  \hspace{1cm} (19)

\[
I(t) = \begin{cases} 
\sum_{k=0}^{K} I_1(k) t^a k, & t \in [0, t_1] \\
\sum_{k=0}^{K} I_2(k) (t - t_1)^a k, & t \in [t_1, t_2] \\
\sum_{k=0}^{K} I_M(k) (t - t_{M-1})^a k, & t \in [t_{M-1}, t_M] 
\end{cases}
\]  \hspace{1cm} (20)

\[
R(t) = \begin{cases} 
\sum_{k=0}^{K} R_1(k) t^a k, & t \in [0, t_1] \\
\sum_{k=0}^{K} R_2(k) (t - t_1)^a k, & t \in [t_1, t_2] \\
\sum_{k=0}^{K} R_M(k) (t - t_{M-1})^a k, & t \in [t_{M-1}, t_M] 
\end{cases}
\]  \hspace{1cm} (21)

\[
C(t) = \begin{cases} 
\sum_{k=0}^{K} C_1(k) t^a k, & t \in [0, t_1] \\
\sum_{k=0}^{K} C_2(k) (t - t_1)^a k, & t \in [t_1, t_2] \\
\sum_{k=0}^{K} C_M(k) (t - t_{M-1})^a k, & t \in [t_{M-1}, t_M] 
\end{cases}
\]  \hspace{1cm} (22)
Here \( S_j(k), I_j(k), R_j(k), C_j(k) \) and \( N_j(k) \) for \( j = 1, 2, \ldots, M \) satisfy the following recurrence relations

\[
S_j(k + 1) = \frac{\Gamma(a k + 1)}{\Gamma((a k + 1) + 1)} \left( \mu (1 - S_j(k)) - \beta \sum_{i=0}^{k} S_j(k-i) I_j(i) + \gamma C_j(k) \right),
\]

\[
I_j(k + 1) = \frac{\Gamma(a k + 1)}{\Gamma((a k + 1) + 1)} \left( \beta \sum_{i=0}^{k} S_j(k-i) I_j(i) + \sigma \beta \sum_{i=0}^{k} C_j(k-i) I_j(i) - (\mu + \delta) I_j(k) \right),
\]

\[
R_j(k + 1) = \frac{\Gamma(a k + 1)}{\Gamma((a k + 1) + 1)} \left( (1 - \sigma) \beta \sum_{i=0}^{k} C_j(k-i) I_j(i) + \theta I_j(k) - (\mu + \delta) R_j(k) \right),
\]

\[
C_j(k + 1) = \frac{\Gamma(a k + 1)}{\Gamma((a k + 1) + 1)} \left( \delta R_j(k) - \beta \sum_{i=0}^{k} C_j(k-i) I_j(i) - (\mu + \gamma) C_j(k) \right),
\]

\[
N_j(k + 1) = \frac{\Gamma(a k + 1)}{\Gamma((a k + 1) + 1)} \left( \mu - \mu N_j(k) \right).
\]

With the initial conditions \( S_j(0) = S_{j-1}(0), I_j(0) = I_{j-1}(0), R_j(0) = R_{j-1}(0), C_j(0) = C_{j-1}(0), \) and \( N_j(0) = N_{j-1}(0) \). Finally, we start with initial conditions \( S(0) = S_0, I(0) = I_0, R(0) = R_0, C(0) = C_0, \) and \( N(0) = N_0 \). And use the recurrence relation given in the above system, we can obtained the multi-step generalized differential transform solution given in (19)-(23).

5. Numerical Method and Simulation

We solve analytically the equations (3)-(7) with initial conditions by using the multi-step generalized differential transform method (MSGDTM). We also numerically solve by nonstandard numerical method and forth-order Runge-Kutta method for numerical results.
For numerical simulation we use a set of parameters given in Table 1. To demonstrate the effectiveness of proposed algorithm as an approximate tool for solving the nonlinear system of fractional differential equations (3)-(7) for large time $t$, we apply this algorithm on the interval $[0-30]$.

### Table 1: Parameter values for the numerical simulation

<table>
<thead>
<tr>
<th>Notation</th>
<th>Parameter description</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\mu$</td>
<td>Natural death rate</td>
<td>0.02</td>
</tr>
<tr>
<td>$\beta$</td>
<td>Contact rate of infection</td>
<td>0.03</td>
</tr>
<tr>
<td>$\gamma$</td>
<td>Cross-immune period</td>
<td>0.05</td>
</tr>
<tr>
<td>$\theta$</td>
<td>Infectious period</td>
<td>0.07</td>
</tr>
<tr>
<td>$\delta$</td>
<td>Total immune period</td>
<td>0.01</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>Fraction of the exposed cross-immune individuals who are recruited in a unit time into the infective subpopulation</td>
<td>0.08</td>
</tr>
</tbody>
</table>

From the graphical results in Figs. 1-5, it can be seen that the results obtained using the multi-step generalized differential transform method match the results of the classical Runge–Kutta method very well, which implies that the presented method can predict the behavior of these variables accurately for the region under consideration. Figs. 6–10 show the approximate solutions for $S(t), E(t), I(t), R(t)$ and $N(t)$ obtained for different values of $\alpha$ using the multi-step generalized differential transform method.
Fig 1. Shows the susceptible individuals.

Fig 2. Shows the infected individuals.
Fig. 3. Shows the recovered individuals

Fig 4. Shows the cross-immune individuals
Fig 5. Shows the total time dependent population.

Fig. 6. $s(t)$ versus $t$ (solid line) MSGDTM, (dotted line) Runge-Kutta method.
Fig. 7. $I(t)$ versus $t$: (solid line) MSGDTM, (dotted line) Runge-Kutta method.

Fig. 8. $R(t)$ versus $t$: (solid line) MSGDTM, (dotted line) Runge-Kutta method.
**Fig. 9.** $C(t)$ versus $t$ (solid line) MSGDTM, (dotted line) Runge-Kutta method.

**Fig. 10.** $N(t)$ versus $t$ (solid line) MSGDTM, (dotted line) Runge-Kutta method.
Fig. 11. $S(t)$ versus $t$ (solid line) $\alpha = 1.0$ (dashed line) $\alpha = 0.95$, (dot-dashed line) $\alpha = 0.85$.

Fig. 12. $I(t)$ versus $t$ (solid line) $\alpha = 1.0$ (dashed line) $\alpha = 0.95$, (dot-dashed line) $\alpha = 0.85$. 
Fig. 13. $R(t)$ versus $t$ (solid line) $\alpha = 1.0$ (dashed line) $\alpha = 0.95$ , (dot-dashed line) $\alpha = 0.85$.

Fig. 14. $C(t)$ versus $t$ (solid line) $\alpha = 1.0$ (dashed line) $\alpha = 0.95$ , (dot-dashed line) $\alpha = 0.85$. 
Fig. 15. $N(t)$ versus $t$ (solid line) $\alpha = 1.0$ (dashed line) $\alpha = 0.95$, (dot-dashed line) $\alpha = 0.85$.

6. Conclusion

In this paper, a fractional order system for $SIRC$ (Susceptible-Infected-Recovered-Cross-immune) epidemic model is studied and its approximate solution is presented using the multi-step generalized differential transform method (MSGDTM).

The approximate solution obtained by multi-step generalized differential transform method are highly accurate and valid for a long time in the integer case. This method is very applicable and also this is a good approach for the solutions of differential equations of such order.

This tool is the best one for modeling in science and engineering.

REFERENCES

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