



Numerical solution for second order Fuzzy Differential Equation by two methods

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Abstract: In this paper, two different methods were used to solve the 2nd fuzzy differential equation. There, Picard's and Runge-Kutta methods. The two methods were applied on MATLAB program to obtain comparative results. Both methods have shown numerical and graphical results which are close to the practical theorem (the exact solution) approximate solution with practical theorem (the exact solution).

Keywords: Fuzzy differential equation, fuzzy numbers, Runge-Kutta method, Picard's iterative method.

I. Introduction

The fuzzy set was first introduced by Zadeh [1]. The theory of blurry groups is one of the important theories currently used in the processing of ambiguous information in mathematical models and is a powerful tool for modeling uncertainty. Therefore, for such mathematical modelling, using fuzzy differential equations are necessary. Important element of fuzzy differential equation was fuzzy derivative (see [2]). Many authors studied the solve of the fuzzy differential equation by numerical method in the same and different method as [3], [4], [5], [7], [8], [9], [10].

In this paper, firstly, the fuzzy differential equation has been defined. Fuzzy differential equations play an important part in the biology, physics, and engineering as well as possible among other fields of science.

After that we used two methods to solve the 2nd fuzzy differential equation, the first method is Runge-Kutta method and discuss the Picard successive approximation method. Then, we explained the two methods by using an example and found the exact and approximate solutions by applying the MATLAB program. Eventually, the obtained results for the exact and approximation solutions were tabulated and discussed.

After that consider the numerical algorithm an approximation method for finding power series solution fuzzy differential equation. $\dot{y} = f(t, y, \hat{y})$.

II .Preliminaries

In this part, we show some definitions , and impotent notations preliminaries, which are used in this [2],[3],[11].

Definition2.1[11]:

A fuzzy number is a map $u: R \rightarrow [0, 1]$ which satisfies

- a. u is upper semicontinuous
- b. $u(x) = 0$ outside some interval $[c, d] \subset R$.
- c. There real numbers a, b such that $c \leq a \leq b \leq d$ where
 1. $u(x)$ is monotonic increasing on $[c, a]$,
 2. $u(x)$ is monotonic decreasing on $[b, d]$,
 3. $u(x) = 1$, $a \leq x \leq b$

A set E is the family of fuzzy numbers and arbitrary fuzzy number is exemplify by an ordered pair of functions

Definition2.2. [11] :

The arbitrary fuzzy number inparametric form is represented by ordered pair functions

$$[u]_r = [\underline{u}(r), \bar{u}(r)] , 0 \leq r \leq 1$$

That is satisfy the following:

1. $\underline{u}(r)$ a bounded left continuous nondecreasing function on $[0,1]$.
2. $\bar{u}(r)$ abounded right continuous nonincreasing function on $[0,1]$.
3. $\underline{u}(r) \leq \bar{u}(r)$ for all $r \in [0,1]$.

A crisp number α is frugally represented by $\underline{u}(r) = \bar{u}(r) = \alpha$, $0 \leq r \leq 1$. For arbitrary,

$$u = (\underline{u}(r), \bar{u}(r)), v = (\underline{v}(r), \bar{v}(r))$$

and $k \in R$, we define equality, addition and multiplication by k as

- a. $u = v$ if and only if $\underline{u}(r) = \underline{v}(r), \bar{u}(r) = \bar{v}(r)$,
- b. $u + v = (\underline{u}(r) + \underline{v}(r), \bar{u}(r) + \bar{v}(r))$,
- c. $ku = \begin{cases} (k\underline{u}, k\bar{u}), & k \geq 0 \\ (k\bar{u}, k\underline{u}), & k < 0. \end{cases}$

Definition 2.3. [11]:

Let I is a real interval. A mapping $X:I \rightarrow E$ is fuzzy process and its r -level set is denoted by $[x(t)]_r = [x_\downarrow(t;r), x_\uparrow(t;r)]$, $t \in I, r \in (0, 1]$ the derivative $\dot{x}(t)$ of fuzzy process x is defined by $[\dot{x}(t)]_r = [\dot{x}_\downarrow(t;r), \dot{x}_\uparrow(t;r)]$, $t \in I, r \in (0, 1]$.

III. Second Order Fuzzy Differential Equation:

2nd fuzzy initial value differential equation is :

$$\begin{cases} \dot{y} = f(t, y, \dot{y}) & t \in [t_0, T] \\ y(t_0) = y_0, \dot{y}(t_0) = y_0 \end{cases}$$

Such that y is a fuzzy function of $t, f(t, y)$ is a fuzzy function of the scrips variable t and the fuzzy Variable y .

y_0 the fuzzy derivative of y and $y(t_0) = y_0$ is a triangular or a triangular shaped fuzzy number. We denote $y = [\underline{y}, \bar{y}]$. It denote that the r -level set of $y(t)$ for $t \in [t_0, T]$ is

$$[y(t)]_r = [\underline{y}(t;r), \bar{y}(t;r)]$$

$$[\dot{y}(t)]_r = [\underline{\dot{y}}(t;r), \bar{\dot{y}}(t;r)]$$

$$[f(t, y(t))]_r = [\underline{f}(t, y(t);r), \bar{f}(t, y(t);r)]$$

Also:

We write:

$$f(t, y) = [\underline{f}(t, y), \bar{f}(t, y)]$$

We have:

$$\begin{aligned} \underline{\dot{y}}(t;r) &= \underline{f}(t, y(t);r) = F[t, \underline{y}(t;r), \bar{y}(t;r)] \\ \bar{\dot{y}}(t;r) &= \bar{f}(t, y(t);r) = G[t, \underline{y}(t;r), \bar{y}(t;r)] \end{aligned}$$

Also we write

$$\begin{aligned} [y(t_0)]_r &= [\underline{y}(t_0;r), \bar{y}(t_0;r)] \\ [y_0]_r &= [(\underline{y_0}(r), \bar{y_0}(r))] \\ \underline{y}(t_0;r) &= \underline{y_0}(r) \quad \bar{y}(t_0;r) = \bar{y_0} \end{aligned}$$

using the extension precept , have themembership function:

$$f(t, y(t))(s) = \sup\{y(t)(\tau) | s = f(t, \tau)\}, \quad s \in R$$

So Fuzzy number $f(t, y(t))$

$$[f(t, y(t))]_r = [f(t, y(t); r), \bar{f}(t, y(t); r)], \quad r \in [0, 1]$$

Where

$$\begin{aligned} f(t, y(t); r) &= \min \{f(t, u) | u \in [y(t)]_r\} \\ \bar{f}(t, y(t); r) &= \max \{f(t, u) | u \in [y(t)]_r\} \end{aligned}$$

IV. Fourth Order Runge-Kutta Method in fuzzy differentialequation

The form of 2nd fuzzy differential equation is

$$\begin{cases} \dot{y} = f(t, y, \dot{y}) & t \in [t_0, T] \\ y(t_0) = y_0, \dot{y}(t_0) = \dot{y}_0 \end{cases} \quad (1)$$

From (1) can reduced two first order simultaneous fuzzy differential equation as

$$\begin{cases} \dot{y} = f(t, \dot{y}, y) \\ \dot{y} = g(t, \dot{y}, y) & t \in [t_0, T] \\ y(t_0) = y_0, \dot{y}(t_0) = \dot{y}_0 \end{cases}$$

The exact solution will be

$$\begin{aligned} [Y(t_n)]_r &= [Y(t_n; r), \bar{Y}(t_n; r)], \\ [\dot{Y}(t_n)]_r &= [\dot{Y}(t_n; r), \bar{\dot{Y}}(t_n; r)] \end{aligned}$$

The approximation solution is given by

$$\begin{aligned} [y(t_n)]_r &= [y(t_n; r), \bar{y}(t_n; r)] \\ [\dot{y}(t_n)]_r &= [\dot{y}(t_n; r), \bar{\dot{y}}(t_n; r)] \end{aligned}$$

By using fourth order Runge-Kutta method, we have

$$[y(t_n)]_r = [y(t_n; r), \bar{y}(t_n; r)]$$

$$\underline{y}(t_{n+1}; r) = \underline{y}(t_n; r) + \sum_{j=1}^4 w_j k_{j,1}(t_n, y(t_n, r)), \quad \bar{y}(t_{n+1}; r) = \bar{y}(t_n; r) + \sum_{j=1}^4 w_j k_{j,2}(t_n, y(t_n, r))$$

$$\underline{\dot{y}}(t_{n+1}; r) = \underline{\dot{y}}(t_n; r) + \sum_{j=1}^4 w_j l_{j,1}(t_n, y(t_n, r)), \quad \bar{\dot{y}}(t_{n+1}; r) = \bar{\dot{y}}(t_n; r) + \sum_{j=1}^4 w_j l_{j,2}(t_n, y(t_n, r))$$

Where $k_{j,\gamma}, k_{j,\gamma}$ could be defined as follow:

$$k_{1,1}(t_n, y(t_n; r)) = \min \{y(t_n, u, v) | u \in (\underline{y}(t_n; r), \bar{y}(t_n; r)), v \in (\underline{\dot{y}}(t_n; r), \bar{\dot{y}}(t_n; r))\}$$

$$k_{1,2}(t_n, y(t_n; r)) = \max \{y(t_n, u, v) | u \in (\underline{y}(t_n; r), \bar{y}(t_n; r)), v \in (\underline{\dot{y}}(t_n; r), \bar{\dot{y}}(t_n; r))\}$$

$$k_{2,1}(t_n, y(t_n; r)) = \min \{y(t_n + \frac{h}{2}, u) | u \in (p_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$k_{2,1}(t_n, y(t_n; r)) = \max \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$k_{3,1}(t_n, y(t_n; r)) = \min \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$k_{3,1}(t_n, y(t_n; r)) = \max \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$k_{4,1}(t_n, y(t_n; r)) = \min \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$k_{4,1}(t_n, y(t_n; r)) = \max \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in (q_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}$$

$$l_{1,1}(t_n, y(t_n; r)) = \min \{g(t_n, u, v) | u \in (\underline{y}(t_n; r), \bar{y}(t_n; r)), v \in (\underline{\dot{y}}(t_n; r), \bar{\dot{y}}(t_n; r))\}$$

$$l_{1,2}(t_n, y(t_n; r)) = \max \{g(t_n, u, v) | u \in (\underline{y}(t_n; r), \bar{y}(t_n; r)), v \in (\underline{\dot{y}}(t_n; r), \bar{\dot{y}}(t_n; r))\}$$

$$\begin{aligned}
 l_{2,1}(t_n, y(t_n; r)) &= \min \{g(t_n + \frac{h}{2}, u) | u \in (p_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \\
 &\quad \in (q_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\} \\
 l_{2,1}(t_n, y(t_n; r)) &= \max \{g(t_n + \frac{h}{2}, u, v) | u \in (p_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \\
 &\quad \in (q_{1,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{1,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\} \\
 l_{3,1}(t_n, y(t_n; r)) &= \min \{g(t_n + \frac{h}{2}, u, v) | u \in (p_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \\
 &\quad \in (q_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\} \\
 l_{3,1}(t_n, y(t_n; r)) &= \max \{g(t_n + \frac{h}{2}, u, v) | u \in (p_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \\
 &\quad \in (q_{2,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{2,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\} \\
 l_{4,1}(t_n, y(t_n; r)) &= \min \{g(t_n + \frac{h}{2}, u, v) | u \in (p_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \\
 &\quad \in (q_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\} \\
 l_{4,1}(t_n, y(t_n; r)) &= \max \{y(t_n + \frac{h}{2}, u, v) | u \in (p_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), p_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r))), v \in \\
 &\quad v \in (q_{3,1}(t_n; y(t_n, r), \dot{y}(t_n, r)), q_{3,2}(t_n; y(t_n, r), \dot{y}(t_n, r)))\}
 \end{aligned}$$

Where:

$$\begin{aligned}
 p_{\lambda, \gamma}(t_n; y(t_n, r)) &= \underline{y}(t_n, r) + \frac{h}{\gamma} k_{\lambda, \gamma}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 p_{\lambda, \gamma}(t_n; y(t_n, r)) &= \bar{y}(t_n, r) + \frac{h}{\gamma} k_{\lambda, \gamma}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 p_{\gamma, \lambda}(t_n; y(t_n, r)) &= \underline{y}(t_n, r) + \frac{h}{\gamma} k_{\gamma, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 p_{\gamma, \lambda}(t_n; y(t_n, r)) &= \bar{y}(t_n, r) + \frac{h}{\gamma} k_{\gamma, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 p_{\tau, \lambda}(t_n; y(t_n, r)) &= \underline{y}(t_n, r) + \frac{h}{\gamma} k_{\tau, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 p_{\tau, \lambda}(t_n; y(t_n, r)) &= \bar{y}(t_n, r) + \frac{h}{\gamma} k_{\tau, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 q_{\lambda, \gamma}(t_n; y(t_n, r)) &= \underline{\dot{y}}(t_n, r) + \frac{h}{\gamma} l_{\lambda, \gamma}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 q_{\lambda, \gamma}(t_n; y(t_n, r)) &= \bar{\dot{y}}(t_n, r) + \frac{h}{\gamma} l_{\lambda, \gamma}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 q_{\gamma, \lambda}(t_n; y(t_n, r)) &= \underline{\dot{y}}(t_n, r) + \frac{h}{\gamma} l_{\gamma, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 q_{\gamma, \lambda}(t_n; y(t_n, r)) &= \bar{\dot{y}}(t_n, r) + \frac{h}{\gamma} l_{\gamma, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r)) \\
 q_{\tau, \lambda}(t_n; y(t_n, r)) &= \underline{\dot{y}}(t_n, r) + \frac{h}{\gamma} l_{\tau, \lambda}(t_n, y(t_n; r), \dot{y}(t_n, r))
 \end{aligned}$$

$$q_{r,r}(t_n; y(t_n, r)) = \bar{y}(t_n, r) + \frac{h}{r} l_{r,r}(t_n, y(t_n; r), \dot{y}(t_n, r))$$

Now using the initial condition, we compute:

$$\underline{y}(t_{n+1}; r) = \underline{y}(t_n; r) + \frac{h}{6}(k_{1,1}(t_n, y(t_n; r)) + 2k_{2,1}(t_n, y(t_n; r)) + 2k_{3,1}(t_n, y(t_n; r)) + k_{4,1}(t_n, y(t_n; r)))$$

$$\bar{y}(t_{n+1}; r) = \bar{y}(t_n; r) + \frac{h}{6}(k_{1,2}(t_n, y(t_n; r)) + 2k_{2,2}(t_n, y(t_n; r)) + 2k_{3,2}(t_n, y(t_n; r)) + k_{4,2}(t_n, y(t_n; r)))$$

$$\underline{y}'(t_{n+1}; r) = \underline{y}'(t_n; r) + \frac{h}{6}(l_{1,1}(t_n, y(t_n; r)) + 2l_{2,1}(t_n, y(t_n; r)) + 2l_{3,1}(t_n, y(t_n; r)) + l_{4,1}(t_n, y(t_n; r)))$$

$$\bar{y}'(t_{n+1}; r) = \bar{y}'(t_n; r) + \frac{h}{6}(k_{1,2}(t_n, y(t_n; r)) + 2k_{2,2}(t_n, y(t_n; r)) + 2k_{3,2}(t_n, y(t_n; r)) + k_{4,2}(t_n, y(t_n; r)))$$

The solution at $t_n, n \leq N$ and $a = t_0 \leq t_1 \leq t_2 \leq \dots \leq t_n = b$, and $h = \frac{b-a}{N} = t_{n+1} - t_n$.

Example (1.1):

Let a fuzzy initial value problem

$$\dot{\hat{y}} - \xi \dot{y} + \xi y = \cdot, \quad t \in [0, 1]$$

$$y(0) = (2 + r, 4 - r)$$

$$\dot{y}(0) = (5 + r, 7 - r)$$

Solve:

Can be reduced the 2nd fuzzy differential equation to two first order fuzzy differential

$$y_1 = y, \quad \dot{y}_1 = \dot{y} = y_r, \quad \dot{y}_2 = \dot{\hat{y}} = \xi \dot{y} - \xi y$$

$$\text{Then } y_r = \dot{y} = f, \quad \dot{\hat{y}} = g = \xi y_r - \xi y_1$$

The exact solution is as follow

$$\underline{y}(t) = (r + 2)e^{rt} + (1 - r)te^{rt}$$

$$\underline{y}(t) = (\xi - r)e^{\xi t} + (r - 1)te^{\xi t}$$

Numerical results

We used MATLAB software in all the calculations which were done in this section.

R	Y lower	Upper	\tilde{Y} lower	Upper
0	3.097279125469690	5.467304090157996	6.857961009099550	9.345805422155820
0.1	3.219392040370772	5.347394508590248	7.015806563085697	9.249946534836342
0.2	3.341504955271854	5.227484927022498	7.173652117071844	9.154087647516860
0.3	3.463617870172936	5.107575345454750	7.331497671057989	9.058228760197380
0.4	3.585730785074018	4.987665763887001	7.489343225044137	8.962369872877899
0.5	3.707843699975100	4.867772848985919	7.647188779030285	8.866644318891753
0.6	3.829956614876182	4.747899934084837	7.805034333016431	8.771078764905607
0.7	3.951189529777264	4.628027019183755	7.943679887002579	8.675513210919458
0.8	4.071762444678345	4.508154104282674	8.067925440988724	8.579947656933314
0.9	4.192335359579428	4.388281189381592	8.192170994974871	8.484382102947166
1	4.312908274480510	4.568408274480509	8.316416548961020	8.388816548961020

Table1: represent approximate solution to Rung-Kutta method

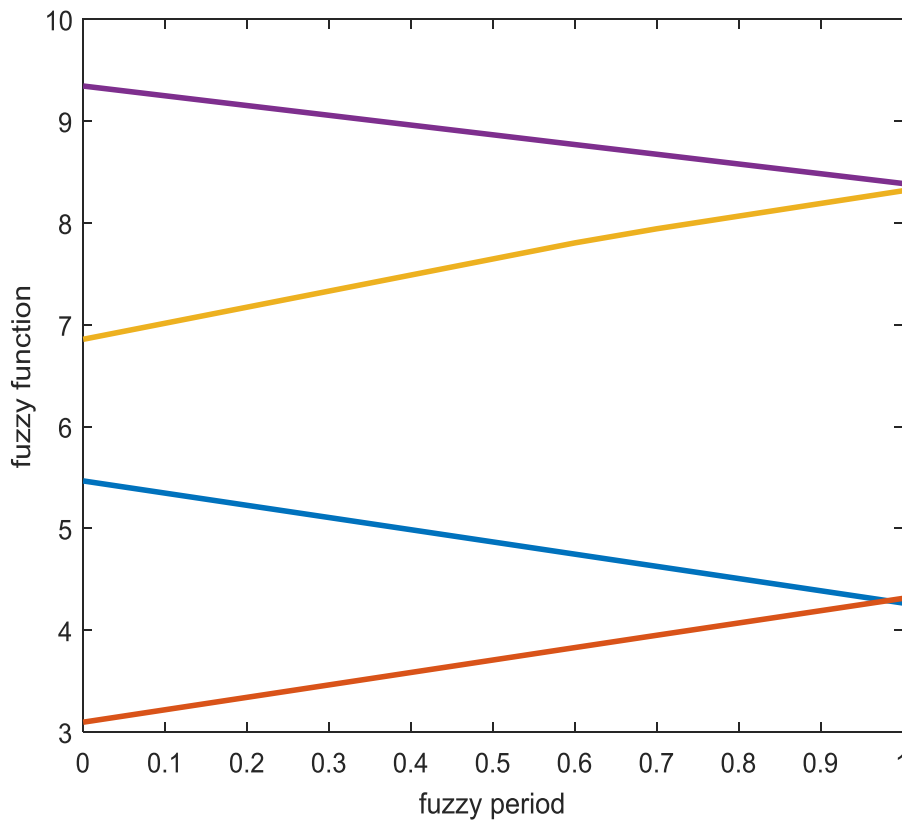


Figure 1: represent Rung-kutta method

V. Fourth Order Picard’s iterative Method in fuzzy differential equation

By using fourth order by Picard’s iterative Method of second order fuzzy differential equation

$$\begin{aligned} \underline{y}_n(t_{n+1}; r) &= \underline{y}(t_n; r) + \int_{t_0}^t \underline{f}(r, s, y_{n-1}(r, s)) ds \\ \overline{y}_n(t_{n+1}; r) &= \overline{y}(t_n; r) + \int_{t_0}^t \overline{f}(r, s, y_{n-1}(r, s)) ds \\ \underline{y}'_n(t_{n+1}; r) &= \underline{y}'(t_n; r) + \int_{t_0}^t \underline{g}(r, s, y_{n-1}(r, s)) ds \\ \overline{y}'_n(t_{n+1}; r) &= \overline{y}'(t_n; r) + \int_{t_0}^t \overline{g}(r, s, y_{n-1}(r, s)) ds \end{aligned}$$

By using the example (1.1) in this method where $\alpha \leq r \leq 1$, $t_0=0$, $n=1,2,3,4$ and the two function f, g is

$$y_\alpha = y, \quad \underline{y}'_\alpha = \underline{y}' = y'_\alpha, \quad \underline{y}''_\alpha = \underline{y}'' = \xi \underline{y}' - \xi y$$

$$\text{Then } y'_\alpha = \underline{y}', \quad \underline{y}''_\alpha = f, \quad \underline{y}'''_\alpha = g = \xi y'_\alpha - \xi y_\alpha$$

$$\begin{aligned} \underline{y}_1(t_{n+1}; r) &= \underline{y}(t_n; r) + \int_0^t \underline{f}(r, s, y_{n-1}(r, s)) ds \\ \underline{y}_2(t_{n+1}; r) &= \underline{y}(t_n; r) + \int_0^t \underline{y}_1(t_{n+1}; r) ds \\ \underline{y}_3(t_{n+1}; r) &= \underline{y}(t_n; r) + \int_0^t \underline{y}_2(t_{n+1}; r) ds \\ \underline{y}_4(t_{n+1}; r) &= \underline{y}(t_n; r) + \int_0^t \underline{y}_3(t_{n+1}; r) ds \\ \overline{y}_1(t_{n+1}; r) &= \overline{y}(t_n; r) + \int_0^t \overline{f}(r, s, y_{n-1}(r, s)) ds \\ \overline{y}_2(t_{n+1}; r) &= \overline{y}(t_n; r) + \int_0^t \overline{y}_1(t_{n+1}; r) ds \\ \overline{y}_3(t_{n+1}; r) &= \overline{y}(t_n; r) + \int_0^t \overline{y}_2(t_{n+1}; r) ds \\ \overline{y}_4(t_{n+1}; r) &= \overline{y}(t_n; r) + \int_0^t \overline{y}_3(t_{n+1}; r) ds \\ \underline{y}'_1(t_{n+1}; r) &= \underline{y}'(t_n; r) + \int_0^t \underline{g}(r, s, y_{n-1}(r, s)) ds \\ \underline{y}'_2(t_{n+1}; r) &= \underline{y}'(t_n; r) + \int_0^t \underline{y}'_1(t_{n+1}; r) ds \\ \underline{y}'_3(t_{n+1}; r) &= \underline{y}'(t_n; r) + \int_0^t \underline{y}'_2(t_{n+1}; r) ds \end{aligned}$$

we obtain

$$\underline{y}_4(t_{n+1}; r) = \underline{y}(t_n; r) + \int_0^t \underline{y}_3(t_{n+1}; r) ds$$

$$\overline{y}_1(t_{n+1}; r) = \overline{y}(t_n; r) + \int_{t_0}^t \underline{g}(r, s, y_{n-1}(r, s)) ds$$

$$\overline{y}_2(t_{n+1}; r) = \overline{y}(t_n; r) + \int_0^t \overline{y}_1(t_{n+1}; r) ds$$

$$\overline{y}_3(t_{n+1}; r) = \overline{y}(t_n; r) + \int_0^t \overline{y}_2(t_{n+1}; r) ds$$

$$\overline{y}_4(t_{n+1}; r) = \overline{y}(t_n; r) + \int_0^t \overline{y}_3(t_{n+1}; r) ds$$

Numerical results

We used MATLAB software in all the calculations, which were done in this section

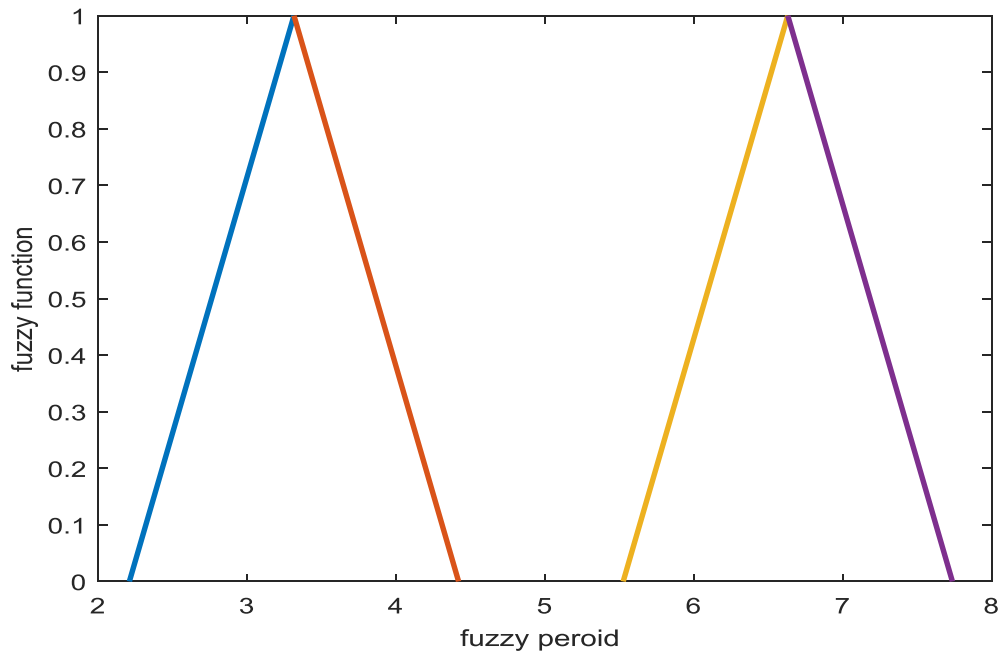
R	Y_nLOW	Y_nUPP	Ŷ_nLOW	Ŷ_nUPP
0	2.2104	4.4207	5.5259	7.7370
.1	2.3209	4.3102	5.6262	7.6264
.2	2.4314	4.1997	5.7265	7.5158
.3	2.5419	4.0891	5.8268	7.4052
.4	2.6524	3.9786	5.9271	7.2946
.5	2.7629	3.8681	6.0274	7.1840
.6	2.8735	3.7576	6.1277	7.0734
.7	2.9840	3.6471	6.2281	6.9628
.8	3.0945	3.5365	6.3283	6.8522
.9	3.2050	3.4260	6.4286	6.7416
1	3.3155	3.3155	6.5289	6.6310

Table 2



r	Y_{nLOW}	Y_{nUPP}	\tilde{Y}_{nLOW}	\tilde{Y}_{nUPP}
0	2.2104	4.4207	5.5259	7.7363
0.1	2.3209	4.3102	5.6364	7.6257
0.2	2.4314	4.1997	5.7470	7.5152
0.3	2.5419	4.0891	5.8575	7.4047
0.4	2.6524	3.9786	5.9680	7.2942
0.5	2.7629	3.8681	6.0785	7.1836
0.6	2.8735	3.7576	6.1890	7.0731
0.7	2.9840	3.6471	6.2995	6.9626
0.8	3.0945	3.5366	6.4100	6.8521
0.9	3.2050	3.4260	6.5205	6.7415
1	3.3155	3.3155	6.6311	6.6310

Table 3



Conclusion:

Form the observation of the approximate and practical result, we observed efficiency of using both methods .However, Picard iteration method was more accurate and less ration of error between the approximate and theoretical



solution .This is clear as explained by table 3, where the theoretical solution is compared with table 2 obviously, Picard iteration method has shown very simple differences between the result.

Fuzzy differential equations serve as mathematical models for many exciting real-world problems, not only in science and technology but also in such diverse fields as population models ,civil engineering, etc

REFERENCES

- [1] L. A. Zadeh, "Fuzzy sets", *Information and Control*, Vol.8, pp. 338-353, (1965).
- [2] O. Kaleva, *Fuzzy differential equations*, *Fuzzy Sets and Systems*, 24, 301-317 (1987).
- [3] S. Abbasbandy and T. Allah Viranloo, "Numerical solution of fuzzy differential equation," *Mathematical & Computational Applications*, vol. 7, no. 1, pp. 41–52, (2002).
- [4] S. Abbasbandy, T. A. Viranloo, O. L´opez-Pouso, and J. Nieto, "Numerical methods for fuzzy differential inclusions," *Computers & Mathematics with Applications*, vol. 48, no. 10-11, pp. 1633–1641, (2004).
- [5] T. Allahviranloo, N. Ahmady, and E. Ahmady, "Numerical solution of fuzzy differential equations by predictor-corrector method," *Information Sciences*, vol. 177, no. 7, pp. 1633–1647, (2007).
- [6] T. Allahviranloo, E. Ahmady, and N. Ahmady, "nth-order fuzzy linear differential equations," *Information Sciences*, vol. 178, no.5, pp. 1309–1324, (2008).
- [7] N. SULTANA and A.F.M. Khodadad KHAN, " Power Series Solution of Fuzzy Differential Equations Using Picard’s Method, *Journal of Applied Computer Science & Mathematics*, no. 8 (4) (2010).
- [8] M. Barkhordari Ahmadi and M. Khezerloo, "Fuzzy bivariate Chebyshev method for solving fuzzy Volterra Fredholm integral equations," *International Journal of Industrial Mathematics*, vol. 3, pp. 67–77, (2011).
- [9] B. Ghazanfari and A. Shakerami, "Numerical solutions of fuzzy differential equations by extended Runge-Kutta-like formulae of order 4," *Fuzzy Sets and Systems*, vol. 189, pp. 74–91, (2012).
- [10] T. Jayakumar, D. Maheskumar and K. Kanagarajan, *Numerical Solution of Fuzzy Differential Equations by Runge-Kutta Method of Order Five*, *Applied Mathematical Sciences*, Vol. 6, no. 60, 2989 – 3002, (2012).
- [11] N. Parandin, " Numerical Solution of Fuzzy Differential Equations of 2nd-Order by Runge-Kutta Method, *Journal of Mathematical Extension* Vol. 7, No. 3, 47-62, (2013)