

1-1-2021

Runge-Kutta Fehlberg Method for Solving Linear and Nonlinear Fuzzy Fredholm Integro-Differential Equations

Porpattama Hammachukiattikul

Department of Mathematics, Faculty of Science and Technology, Phuket Rajabhat University, Phuket, Thailand-83000

Bundit Unyong

R. Suresh

Grienggrai Rajchakit

R. Vadivel

See next page for additional authors

Follow this and additional works at: <https://dc.naturalspublishing.com/amis>

Recommended Citation

Hammachukiattikul, Porpattama; Unyong, Bundit; Suresh, R.; Rajchakit, Grienggrai; Vadivel, R.; and Gunasekaran, Nallappan (2021) "Runge-Kutta Fehlberg Method for Solving Linear and Nonlinear Fuzzy Fredholm Integro-Differential Equations," *Applied Mathematics & Information Sciences*: Vol. 15 : Iss. 1 , Article 6.

DOI: <http://dx.doi.org/10.18576/amis/150106>

Available at: <https://dc.naturalspublishing.com/amis/vol15/iss1/6>

This Article is brought to you for free and open access by Natural Sciences Publishing Digital Commons. It has been accepted for inclusion in Applied Mathematics & Information Sciences by an authorized editor of Natural Sciences Publishing Digital Commons. For more information, please contact nsp@natural-s-publishing.com, amisaty@gmail.com, halimaty@gmail.com.

Runge-Kutta Fehlberg Method for Solving Linear and Nonlinear Fuzzy Fredholm Integro-Differential Equations

Authors

Porpattama Hammachukiattikul, Bundit Unyong, R. Suresh, Grienggrai Rajchakit, R. Vadivel, and Nallappan Gunasekaran

Runge-Kutta Fehlberg Method for Solving Linear and Nonlinear Fuzzy Fredholm Integro-Differential Equations

Porpattama Hammachukiattikul¹, Bundit Unyong¹, R. Suresh², Grienggrai Rajchakit^{3,*}, R. Vadivel¹, Nallappan Gunasekaran⁴ and Chee Peng Lim⁵

¹Department of Mathematics, Faculty of Science and Technology, Phuket Rajabhat University, Phuket, Thailand-83000

²Department of Applied Mathematics, Sri Venkateswara College of Engineering, Sriperumbudur, India 602117

³Department of Mathematics, Faculty of Science, Maejo University, Sansai 50290, Chiang Mai, Thailand

⁴Department of Mathematical Sciences, Shibaura Institute of Technology, Saitama 337-8570, Japan

⁵Institute for Intelligent Systems Research and Innovation, Deakin University, Waurn Ponds, VIC 3216, Australia

Received: 11 Sep. 2020, Revised: 2 Dec. 2020, Accepted: 18 Dec. 2020

Published online: 1 Jan. 2021

Abstract: A study on the parametric form of fuzzy numbers is presented in this paper. The Runge-Kutta Fehlberg method is exploited to yield the approximate solution with respect to the second type of fuzzy Fredholm integro-differential equations. Both linear and nonlinear numerical examples are provided in our analysis. The results ascertain the effectiveness and precision of the proposed method.

Keywords: Integro-differential equations, Runge-Kutta method, Fehlberg method, fuzzy

1. Introduction

Real-world dynamic systems are subject to all kinds of uncertainties, such as population growth [1, 2] and [3], tower-bell oscillation [4], friction of sliding surfaces [5], contaminant migration in porous media [6], and a human life cycle [7]. Generally in mathematics, a random variable or a fuzzy set is used for handling uncertainty. In the 1960s, the theory of fuzzy sets was introduced by Zadeh to deal with uncertainty due to imprecision or vagueness, instead of randomness. Then, studies on fuzzy numbers and associated arithmetic operations were conducted by Zadeh [8] and [9], while further enhancements were proposed by Mizumoto and Tanaka [10]. The notion of LR fuzzy numbers was initiated by Dubios and Prade [11], where a computing mechanism for dealing with fuzzy functions was provided.

The study of Integro-differential Equation, on the other hand, has gained growing interest in various physical, biological and engineering sciences [12, 13] and [14]. In the last two decades, many researchers have been into the analytical and numerical methods for the solution of Integro differential equation. Since then, fuzzy differential and integral equations have been rigorously

improved from fuzzy control application perspectives. In modelling uncertainties in dynamical systems, fuzzy integro-differential equations (FIDE) play an important role. Indeed, they have been successfully used in various domains, including engineering, biology, medicine, physics, and economy. The authors in [15] introduced the existence and uniqueness of solutions pertaining to FIDE. The existing results for fuzzy delay integro-differential equations and general fuzzy volterra-fredholm integral equations have been researched by Balachandran and Kanagarajan [16] and [17]. On the other hand, application of fuzzy integral equations together with control problems and fuzzy uncertainties have been modelled by Diamond [18].

In general, except for a few linear and non-linear systems, it is very difficult to get a analytical solution for an FIDE. To solve Integro-differential equations, several numerical methods are available in the literature [19–25] and [26]. Using the variation iteration technique, Abbasbandy and Hashemi [27] solved FIDE. In [28], the numerical solutions for FIDE were provided by analyzing homotopy. Allahviranloo et al. [29] presented a new technique to tackle FIDE with generalized differentiability.

* Corresponding author e-mail: kreangkri@mju.ac.th



The main contribution of this paper can be summarized as follows.

- i) In this paper, we employ the Rung-Kutta Fehlberg technique to yield the numerical solution of fuzzy Fredholm integro-differential equations (FFIDE).
- ii) Both linear and nonlinear numerical examples are provided in our analysis.
- iii) Finally, simulation results and table values are given to show the advantage and errors obtained by Runge-Kutta Fehlberg method.

2. Problem Formulation and Preliminaries

In this section, the fuzzy number definitions and the fundamental notions used in fuzzy operations are presented.

Definition 1. [30, 31] and [32]: A fuzzy subset, u_1 , is called a fuzzy number when the universal set on which φ_{u_1} is defined is the set of all real numbers, \mathbb{R} , and satisfies the following conditions:

- (i) All the β -levels of u_1 are not empty for $0 \leq \beta \leq 1$;
- (ii) all the β -levels of u_1 are closed intervals of \mathbb{R} ;
- (iii) $\text{supp } u_1 = x \in \mathbb{R} : \varphi_{u_1}(x) > 0$ is bounded.

In addition, \hat{E} denotes the set of all fuzzy numbers. Kaleva [33] provided a substitute definition for producing the same \hat{E} . As discussed in [34], this fuzzy number space can be formulated in the Banach space $B = \overline{C}[0, 1] \times \overline{C}[0, 1]$.

Definition 2. [35]: For arbitrary fuzzy numbers $\tilde{u}_1, \tilde{v}_1 \in \hat{E}$, $\beta \in [0, 1]$, the distance (Hausdorff metric) is employed, i.e.,

$$D(u_1, v_1) = \sup_{\beta \in [0,1]} \max\{|\underline{u}_1(\beta) - \underline{v}_1(\beta)|, |\overline{u}_1(\beta) - \overline{v}_1(\beta)|\}.$$

It has been highlighted in [36] that (\hat{E}, D) is a complete metric space, and some well-established properties are as follows:

$$\begin{aligned} D(\tilde{u}_1 \oplus \tilde{w}_1, \tilde{v}_1 \oplus \tilde{w}_1) &= D(\tilde{u}_1, \tilde{v}_1), \forall \tilde{u}_1, \tilde{v}_1, \tilde{w}_1 \in \hat{E}, \\ D(k \odot \tilde{u}_1, k \odot \tilde{v}_1) &= |k|D(\tilde{u}_1, \tilde{v}_1), \forall k \in \mathbb{R}^1, \tilde{u}_1, \tilde{v}_1 \in \hat{E}, \\ D(\tilde{u}_1 \oplus \tilde{v}_1, \tilde{w}_1 \oplus \tilde{e}_1) &\leq \\ D(\tilde{u}_1, \tilde{w}_1) + D(\tilde{v}_1, \tilde{e}_1), &\forall \tilde{u}_1, \tilde{v}_1, \tilde{w}_1, \tilde{e}_1 \in \hat{E}. \end{aligned}$$

Definition 3. [37]: Suppose $g : [a_1, a_2] \rightarrow \hat{E}$ is a fuzzy function. For arbitrary fixed $t_0 \in \mathbb{R}^1$ and $\hat{\epsilon} > 0, \hat{\delta} > 0$, therefore,

$$|t - t_0| < \hat{\delta} \Rightarrow D(g(t), g(t_0)) < \hat{\epsilon}.$$

Then, g is concluded as continuous.

Definition 4. [35] and [37]: Suppose $g : [a_1, a_2] \rightarrow \hat{E}$. For each partition $R = \{t_0, t_1, \dots, t_n\}$ of $[a_1, a_2]$ and for arbitrary $\xi_i \in [t_{i-1}, t_i], 1 \leq i \leq n$, assume

$$P_R = \sum_{i=1}^n g(\xi_i)(t_i - t_{i-1}), \quad \lambda = \max_{1 \leq i \leq n} \{t_i - t_{i-1}\}.$$

Then, the definite integral of $g(t) \in [a_1, a_2]$ is provided as follows:

$$\int_{a_1}^{a_2} g(t)dt = \lim_{\lambda \rightarrow 0} P_R,$$

and its limit exists in metric D .

Given that the fuzzy function $g(t)$ is continuous in metric D , the definite integral exists [35], and also

$$\overline{\left(\int_{a_1}^{a_2} g(t;\beta)dt\right)} = \int_{a_1}^{a_2} \overline{g(t;\beta)}dt, \quad \underline{\left(\int_{a_1}^{a_2} g(t;\beta)dt\right)} = \int_{a_1}^{a_2} \underline{g(t;\beta)}dt.$$

Definition 5. [38]: Suppose x_1 and $y_1 \in \hat{E}$. There exists $z_1 \in \hat{E}$ and $x_1 = y_1 + z_1$, then z_1 is known as the H -difference of x_1 and y_1 , which is represented by $x_1 \ominus y_1$. In addition, if there exists H -difference $\tilde{u}_1 \ominus \tilde{v}_1$ and $\tilde{w}_1 \ominus \tilde{e}_1$, it can be deduced that $D(\tilde{u}_1 \ominus \tilde{v}_1, \tilde{w}_1 \ominus \tilde{e}_1) = D(\tilde{u}_1 \oplus \tilde{e}_1, \tilde{w}_1 \oplus \tilde{v}_1), \forall \tilde{u}_1, \tilde{v}_1, \tilde{w}_1, \tilde{e}_1 \in \hat{E}$.

In this paper, the " \ominus " sign stands always for H -difference and let us remark that $x_1 \ominus y_1 \neq x_1 + (-1)y_1$. In this study, the following definition of differentiability for fuzzy-valued functions introduced in [38] and investigated in [39] is adopted:

Definition 6. Suppose $g : (a_1, a_2) \rightarrow \hat{E}$ and $r_0 \in (a_1, a_2)$. Then, g is strongly generalized H -differentiable at r_0 , there exists an element $g'(r_0) \in \hat{E}$, therefore,

- (1) for all $h > 0$ sufficiently close to 0, there exist $g(r_0 + h) \ominus g(r_0)$, $g(r_0) \ominus g(r_0 - h)$ and the limits (in the metric D) are:

$$\lim_{h \rightarrow 0^+} \frac{g(r_0 + h) \ominus g(r_0)}{h} = \lim_{h \rightarrow 0^+} \frac{g(r_0) \ominus g(r_0 - h)}{h} = g'(r_0),$$

- (2) for all $h < 0$ sufficiently close to 0, there exists $g(r_0) \ominus g(r_0 + h)$, $g(r_0 - h) \ominus g(r_0)$ and the limits (in the metric D) are:

$$\lim_{h \rightarrow 0^+} \frac{g(r_0) \ominus g(r_0 + h)}{h} = \lim_{h \rightarrow 0^+} \frac{g(r_0 - h) \ominus g(r_0)}{h} = g'(r_0).$$

In the special case when g is a fuzzy-valued function, the consecutive results can be obtained.

Lemma 1. [39]: Suppose $g : \mathbb{R}^1 \rightarrow \hat{E}$ is a function, and denote $g(t) = (g(t; \beta), \overline{g}(t; \beta))$, for each $\beta \in [0, 1]$. Then,

- (1) if g is differentiable in the first form (1) in Definition 6, then $\underline{g}(t; \beta)$ and $\overline{g}(t; \beta)$ are differentiable functions, and $g'(t) = (\underline{g}'(t; \beta), \overline{g}'(t; \beta))$

(2) if g is differentiable in the second form (2) in Definition 6, then $\underline{g}(t; \beta)$ and $\overline{g}(t; \beta)$ are differentiable functions, and $g'(t) = (\underline{g}'(t; \beta), \overline{g}'(t; \beta))$.

In [33], we can obtain the key properties of the gH-derivatives in the first form (1), and some of which still hold for the next form (2). In [39], we can obtain the key properties for the next form (2). Note that fuzzy-valued function g is I-differentiable if it fascinates the first form (1) in Definition 6, while g is II-differentiable if it fascinates the second form (2) in Definition 6.

3. Fuzzy Fredholm Integro-Differential Equations : Runge-Kutta Fehlberg Method

The Fredholm integro-differential equations are expressed as follows: [40] and [41]

$$\begin{cases} \theta'(t) = x(t) + \sigma \int_{a_1}^{a_2} q(t,s)\theta(s)ds \\ \theta(t_0) = \theta_0, \end{cases} \quad (1)$$

where $\sigma > 0$, q is an arbitrary given kernel function over the square $a_1 \leq t, s \leq a_2$. Suppose θ denotes the fuzzy function, $x(t)$ noted as given fuzzy function of $t \in [a_1, a_2]$, and θ' represents a fuzzy derivative (in reference to Definition 6) of θ ; this equation only possesses a fuzzy solution. The sufficient conditions for the existing equation of the second kind has been modelled in [42].

Suppose $\theta(t) = [\underline{\theta}(t; \beta), \overline{\theta}(t; \beta)]$ is a fuzzy solution of equation (1). By Definitions 4 and 6, the equivalent model is obtained:

$$\begin{cases} \underline{\theta}'(t) = \underline{x}(t) + \sigma \int_{a_1}^{a_2} \underline{q}(t,s)\underline{\theta}(s)ds, & \underline{\theta}(t_0) = \underline{\theta}_0 \\ \overline{\theta}'(t) = \overline{x}(t) + \sigma \int_{a_1}^{a_2} \overline{q}(t,s)\overline{\theta}(s)ds, & \overline{\theta}(t_0) = \overline{\theta}_0, \end{cases} \quad (2)$$

and it possesses a unique solution $(\underline{\theta}, \overline{\theta}) \in B$, which is a fuzzy function. Specifically, for each s , the pair $[\underline{\theta}(t; \beta), \overline{\theta}(t; \beta)]$ denotes a fuzzy number, therefore, each solution of (1) is a solution of model (2). In, reverse Model (1) and Model (2) are identical.

The parametric form of Model (2) is given by

$$\begin{cases} \underline{\theta}'(t, \beta) = \underline{x}(t, \beta) + \sigma \int_{a_1}^{a_2} \underline{q}(t,s)\underline{\theta}(s, \beta)ds, \\ \underline{\theta}(t_0) = \underline{\theta}_0(\beta) \\ \overline{\theta}'(t, \beta) = \overline{x}(t, \beta) + \sigma \int_{a_1}^{a_2} \overline{q}(t,s)\overline{\theta}(s, \beta)ds, \\ \overline{\theta}(t_0) = \overline{\theta}_0(\beta), \end{cases} \quad (3)$$

for $\beta \in [0, 1]$. Let $q(t, s)$ be continuous in $a_1 \leq t \leq a_2$. For fixed t , $q(t, s)$ changes the sign in finite points as t_i , where

$\theta_i \in [a_1, t_1]$. As an example, let $q(t, s)$ be non-negative over $[a_1, t_1]$ and negative over $[t_1, a_2]$. Then, we have

$$\begin{cases} \underline{\theta}'(t, \beta) = \underline{x}(t, \beta) + \sigma \int_{a_1}^{t_1} q(t,s)\underline{\theta}(s, \beta)ds \\ \quad + \sigma \int_{t_1}^{a_2} q(t,s)\overline{\theta}(s, \beta)ds, & \underline{\theta}(t_0) = \underline{\theta}_0(\beta) \\ \overline{\theta}'(t, \beta) = \overline{x}(t, \beta) + \sigma \int_{a_1}^{t_1} q(t,s)\overline{\theta}(s, \beta)ds \\ \quad + \sigma \int_{t_1}^{a_2} q(t,s)\underline{\theta}(s, \beta)ds, & \overline{\theta}(t_0) = \overline{\theta}_0(\beta). \end{cases} \quad (4)$$

In several cases, yet, the analytical solution for Eq. (3) cannot be available, and a numerical method has to be used. In the interval $[a_1, a_2]$, consider a set of discrete equally spaced grid points $a_1 < t_0 < t_1 < t_2 < \dots < t_N = a_2$ at which two exact solutions $\Theta(t, \beta) = [\underline{\Theta}(t, \beta), \overline{\Theta}(t, \beta)]$ are approximated by $\theta(t, \beta) = [\underline{\theta}(t, \beta), \overline{\theta}(t, \beta)]$, respectively. The grid points at which the solutions are computed are $t_n = t_0 + nh$, $h = \frac{(a_2 - a_1)}{N}$. The exact and approximate solutions at $t_n, 0 < n < N$ are represented by $\Theta_n(\beta)$ and $\theta_n(\beta)$, respectively. Based on the RKF method, the first-order approximation of $\underline{\Theta}(t, \beta), \overline{\Theta}(t, \beta)$ and $\underline{\theta}(t, \beta), \overline{\theta}(t, \beta)$ is achieved as follows:

$$\begin{cases} \underline{\theta}_{n+1}(\beta) = \underline{\theta}_n(\beta) + \sum_{i=1}^6 w_i \underline{l}_i(t_n, [\theta(t_n)]^\beta) \\ \overline{\theta}_{n+1}(\beta) = \overline{\theta}_n(\beta) + \sum_{i=1}^6 w_i \overline{l}_i(t_n, [\theta(t_n)]^\beta), \end{cases} \quad (5)$$

where w_i 's are constants, and

$$l_i(t_n, [\theta(t_n)]^\beta) = [\underline{l}_i(t_n, [\theta(t_n)]^\beta), \overline{l}_i(t_n, [\theta(t_n)]^\beta)],$$

where

$$\begin{aligned} \underline{l}_i(t_n, [\theta(t_n)]^\beta) &= h F(t_n + \beta_i h, [\theta(t_n)]^\beta + \sum_{j=1}^{i-1} \gamma_{ij} \underline{l}_j(t_n, [\theta(t_n)]^\beta)), \\ \overline{l}_i(t_n, [\theta(t_n)]^\beta) &= h G(t_n + \beta_i h, [\theta(t_n)]^\beta + \sum_{j=1}^{i-1} \gamma_{ij} \overline{l}_j(t_n, [\theta(t_n)]^\beta)), \end{aligned}$$



and

$$\begin{aligned} \underline{l}_1(t_n, [\theta(t_n)]^\beta) &= \min \{ h F [t_n, \underline{\theta}_n(\beta), \bar{\theta}_n(\beta)] \}, \\ \underline{l}_2(t_n, [\theta(t_n)]^\beta) &= \min \left\{ h F \left[t_n + \frac{h}{4}, [\theta(t_n)]^\beta + \frac{1}{4} \underline{l}_1(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \underline{l}_3(t_n, [\theta(t_n)]^\beta) &= \min \left\{ h F \left[t_n + \frac{3h}{8}, [\theta(t_n)]^\beta + \frac{3}{32} \underline{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + \frac{9}{32} \underline{l}_2(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \underline{l}_4(t_n, [\theta(t_n)]^\beta) &= \min \left\{ h F \left[t_n + \frac{12h}{13}, [\theta(t_n)]^\beta + \frac{1932}{2197} \underline{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. - \frac{7200}{2197} \underline{l}_2(t, [\theta(t_n)]^\beta) + \frac{7296}{2197} \underline{l}_3(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \underline{l}_5(t_n, [\theta(t_n)]^\beta) &= \min \left\{ h F \left[t_n + h, [\theta(t_n)]^\beta + \frac{439}{216} \underline{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. - 8 \underline{l}_2(t, [\theta(t_n)]^\beta) + \frac{3680}{513} \underline{l}_3(t, [\theta(t_n)]^\beta) - \frac{845}{4104} \underline{l}_4(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \underline{l}_6(t_n, [\theta(t_n)]^\beta) &= \min \left\{ h F \left[t_n + \frac{h}{2}, [\theta(t_n)]^\beta - \frac{8}{27} \underline{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + 2 \underline{l}_2(t, [\theta(t_n)]^\beta) - \frac{3544}{2565} \underline{l}_3(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + \frac{1859}{4104} \underline{l}_4(t, [\theta(t_n)]^\beta) - \frac{11}{40} \underline{l}_5(t, [\theta(t_n)]^\beta) \right] \right\}. \end{aligned}$$

And

$$\begin{aligned} \bar{l}_1(t_n, [\theta(t_n)]^\beta) &= \max \{ h G [t_n, \underline{\theta}_n(\beta), \bar{\theta}_n(\beta)] \}, \\ \bar{l}_2(t_n, [\theta(t_n)]^\beta) &= \max \left\{ h G \left[t_n + \frac{h}{4}, [\theta(t_n)]^\beta + \frac{1}{4} \bar{l}_1(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \bar{l}_3(t_n, [\theta(t_n)]^\beta) &= \max \left\{ h G \left[t_n + \frac{3h}{8}, [\theta(t_n)]^\beta + \frac{3}{32} \bar{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + \frac{9}{32} \bar{l}_2(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \bar{l}_4(t_n, [\theta(t_n)]^\beta) &= \max \left\{ h G \left[t_n + \frac{12h}{13}, [\theta(t_n)]^\beta \right. \right. \\ &\quad \left. \left. + \frac{1932}{2197} \bar{l}_1(t, [\theta(t_n)]^\beta) - \frac{7200}{2197} \bar{l}_2(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + \frac{7296}{2197} \bar{l}_3(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \bar{l}_5(t_n, [\theta(t_n)]^\beta) &= \max \left\{ h G \left[t_n + h, [\theta(t_n)]^\beta + \frac{439}{216} \bar{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. - 8 \bar{l}_2(t, [\theta(t_n)]^\beta) + \frac{3680}{513} \bar{l}_3(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. - \frac{845}{4104} \bar{l}_4(t, [\theta(t_n)]^\beta) \right] \right\}, \\ \bar{l}_6(t_n, [\theta(t_n)]^\beta) &= \max \left\{ h G \left[t_n + \frac{h}{2}, [\theta(t_n)]^\beta - \frac{8}{27} \bar{l}_1(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + 2 \bar{l}_2(t, [\theta(t_n)]^\beta) - \frac{3544}{2565} \bar{l}_3(t, [\theta(t_n)]^\beta) \right. \right. \\ &\quad \left. \left. + \frac{1859}{4104} \bar{l}_4(t, [\theta(t_n)]^\beta) - \frac{11}{40} \bar{l}_5(t, [\theta(t_n)]^\beta) \right] \right\}. \end{aligned}$$

Define

$$\begin{cases} F[t_n, [\theta(t_n)]^\beta] = \frac{16}{135} \underline{l}_1(t_n, [\theta(t_n)]^\beta) + \frac{6656}{12825} \underline{l}_3(t_n, [\theta(t_n)]^\beta) \\ \quad + \frac{28561}{56430} \underline{l}_4(t_n, [\theta(t_n)]^\beta) - \frac{9}{50} \underline{l}_5(t_n, [\theta(t_n)]^\beta) \\ \quad + \frac{2}{55} \underline{l}_6(t_n, [\theta(t_n)]^\beta) \\ G[t_n, [\theta(t_n)]^\beta] = \frac{16}{135} \bar{l}_1(t_n, [\theta(t_n)]^\beta) + \frac{6656}{12825} \bar{l}_3(t_n, [\theta(t_n)]^\beta) \\ \quad + \frac{28561}{56430} \bar{l}_4(t_n, [\theta(t_n)]^\beta) - \frac{9}{50} \bar{l}_5(t_n, [\theta(t_n)]^\beta) \\ \quad + \frac{2}{55} \bar{l}_6(t_n, [\theta(t_n)]^\beta) \end{cases}$$

From the above equations, we have

$$\begin{cases} \underline{\theta}_{n+1}(\beta) = \underline{\theta}_n(\beta) + F[t_n, \underline{\theta}_n(\beta), \bar{\theta}_n(\beta)] \\ \bar{\theta}_{n+1}(\beta) = \bar{\theta}_n(\beta) + G[t_n, \underline{\theta}_n(\beta), \bar{\theta}_n(\beta)] \\ \underline{\theta}_0(\beta) = \underline{\theta}_0(\beta) \\ \bar{\theta}_0(\beta) = \bar{\theta}_0(\beta) \end{cases} \quad (6)$$

4. Numerical Examples

In this section, in view of the conditions acquired in the previous section, we present some simulation examples to demonstrate the adequacy of the proposed methods and the merits of our approach.

Example 1. Let us consider the following fuzzy integro-differential equation:

$$\begin{cases} \theta'(t) = [0.96 + 0.04\beta, 1.01 - 0.01\beta] [e^t - \frac{t^2}{2}] + \frac{t^2}{2} \int_0^1 s\theta(s)ds \\ \theta(0; \beta) = [0.96 + 0.04\beta, 1.01 - 0.01\beta], \quad 0 \leq \beta \leq 1, \quad 0 \leq t, s \leq 1. \end{cases}$$

The exact solution is given by

$$\Theta(t; \beta) = [(0.96 + 0.04\beta)e^t, (1.01 - 0.01\beta)e^t].$$

The approximate solution, by using the Runge-Kutta Fehlberg method, is given by

$$\begin{cases} \underline{\theta}_{n+1}^\beta = \underline{\theta}_n^\beta [1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24} + \frac{h^5}{120} + \frac{h^6}{2080}] \\ \bar{\theta}_{n+1}^\beta = \bar{\theta}_n^\beta [1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24} + \frac{h^5}{120} + \frac{h^6}{2080}] \\ \underline{\theta}_0^\beta = \underline{\theta}_0 \\ \bar{\theta}_0^\beta = \bar{\theta}_0. \end{cases}$$

Table 1 and Fig. 1 depict a variation among the exact and the estimated solutions at t = 1.

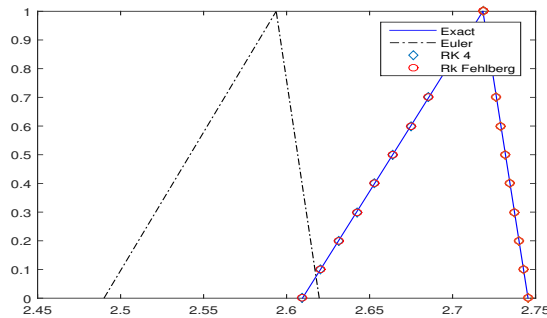


Fig. 1: At t=1 in Ex. 1

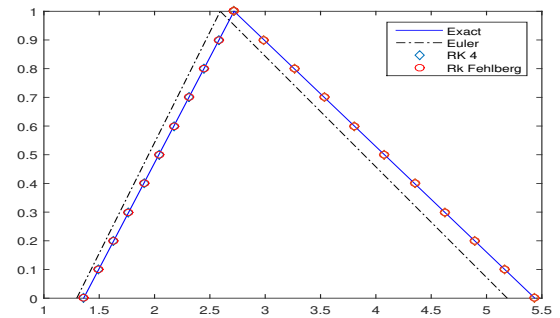


Fig. 3: At t=1 in Ex. 2

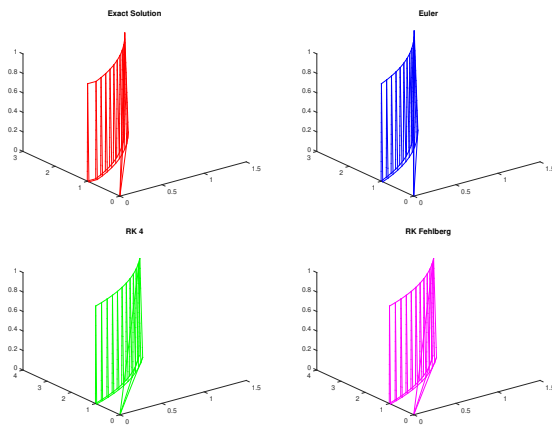


Fig. 2: At t=0.1 in Ex. 1

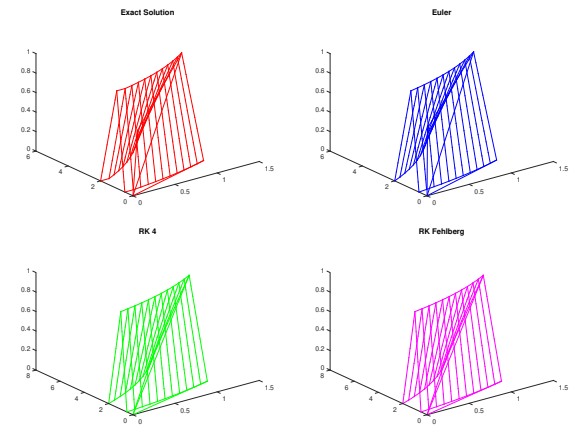


Fig. 4: At t=0.1 in Ex. 2

Example 2. Consider the following fuzzy integro-differential equation:

$$\begin{cases} \theta'(t) = [0.5 + 0.5\beta, 2 - \beta][e^t + (\frac{1 - e^2}{2})t] + \int_0^1 t\theta^2(s)ds \\ \theta(0; \beta) = [0.5 + 0.5\beta, 2 - \beta], \quad 0 \leq \beta \leq 1, \quad 0 \leq t, s \leq 1. \end{cases}$$

The exact solution is given by

$$\Theta(t; \beta) = [(0.5 + 0.5\beta)e^t, (2 - \beta)e^t].$$

The approximate solution, by using RKF method, is given by

$$\begin{cases} \underline{\theta}_{n+1}^\beta = \underline{\theta}_n^\beta [1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24} + \frac{h^5}{120} + \frac{h^6}{2080}] \\ \bar{\theta}_{n+1}^\beta = \bar{\theta}_n^\beta [1 + h + \frac{h^2}{2} + \frac{h^3}{6} + \frac{h^4}{24} + \frac{h^5}{120} + \frac{h^6}{2080}] \\ \underline{\theta}_0^\beta = \underline{\theta}_0 \\ \bar{\theta}_0^\beta = \bar{\theta}_0. \end{cases}$$

Table 2 and Fig. 3 depict a comparison between the exact and the approximate solutions at $t = 1$. In addition, evolution of curves are represented in Figs. 2 and 4 with $t = 0.1$, respectively. From Tables 1 and 2, it can be observed that the errors obtained by the proposed method are better than those from Euler and R-K methods of order four.

Remark 1. To the best of the authors knowledge, very few investigations on solving Fuzzy Fredholm integro-differential equations using numerical techniques [19, 26] and [32]. Therefore, in this paper, we study the Runge-Kutta Fehlberg method for solving linear and nonlinear fuzzy Fredholm integro-differential equations, as summarized in Section 3.



Table 1: Error analysis in Ex. 1

β	Exact solution		Error in Euler method		Error in RK four		Error in RK Fehlberg	
	$\underline{Q}(t, \beta)$	$\overline{Q}(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$
0	2.609550555	2.745454547	0.1196	0.1258	0.2001e-5	0.2105e-5	0.2192e-7	0.2306e-7
0.1	2.620423683	2.742746365	0.1201	0.1257	0.2009e-5	0.2103e-5	0.2201e-7	0.2304e-7
0.2	2.631296810	2.740028083	0.1206	0.1255	0.2018e-5	0.2101e-5	0.2210e-7	0.2301e-7
0.3	2.642169937	2.737309801	0.1211	0.1254	0.2026e-5	0.2099e-5	0.2219e-7	0.2299e-7
0.4	2.653043065	2.734591519	0.1216	0.1253	0.2034e-5	0.2097e-5	0.2228e-7	0.2297e-7
0.5	2.663916192	2.731873238	0.1220	0.1252	0.2043e-5	0.2095e-5	0.2237e-7	0.2294e-7
0.6	2.674789319	2.729154956	0.1225	0.1250	0.2051e-5	0.2093e-5	0.2247e-7	0.2292e-7
0.7	2.685662447	2.726436674	0.1230	0.1249	0.2059e-5	0.2091e-5	0.2256e-7	0.2290e-7
0.8	2.696535574	2.723718392	0.1235	0.1248	0.2068e-5	0.2088e-5	0.2265e-7	0.2288e-7
0.9	2.707408701	2.72100011	0.1240	0.1247	0.2076e-5	0.2086e-5	0.2274e-7	0.2285e-7
1	2.718281828	2.718281828	0.1245	0.1245	0.2084e-5	0.2084e-5	0.2283e-7	0.2283e-7

Table 2: Error analysis in Ex. 2

β	Exact solution		Error in Euler method		Error in RK four		Error in RK Fehlberg	
	$\underline{Q}(t, \beta)$	$\overline{Q}(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$	$\underline{\theta}_1(t, \beta)$	$\overline{\theta}_1(t, \beta)$
0	1.359140914	5.436563657	0.0623	0.2491	0.1042e-5	0.4169e-5	0.1142e-7	0.4566e-7
0.1	1.495055006	5.164735474	0.0685	0.2366	0.1146e-5	0.3960e-5	0.1256e-7	0.4338e-7
0.2	1.630969097	4.892907291	0.0747	0.2242	0.1251e-5	0.3752e-5	0.1370e-7	0.4109e-7
0.3	1.766883188	4.621079108	0.0810	0.2117	0.1355e-5	0.3543e-5	0.1484e-7	0.3881e-7
0.4	1.902797280	4.349250926	0.0872	0.1993	0.1459e-5	0.3335e-5	0.1598e-7	0.3653e-7
0.5	2.038711371	4.077411743	0.0934	0.1868	0.1563e-5	0.3126e-5	0.1712e-7	0.3425e-7
0.6	2.174625463	3.80559456	0.0996	0.1744	0.1667e-5	0.2918e-5	0.1826e-7	0.3196e-7
0.7	2.310539554	3.533766377	0.1059	0.1619	0.1772e-5	0.2710e-5	0.1941e-7	0.2968e-7
0.8	2.446453646	3.261938194	0.1121	0.1494	0.1876e-5	0.2501e-5	0.2055e-7	0.2740e-7
0.9	2.582367737	2.990110011	0.1183	0.1370	0.1980e-5	0.2293e-5	0.2169e-7	0.2511e-7
1	2.718281828	2.718281828	0.1245	0.1245	0.2084e-5	0.2084e-5	0.2283e-7	0.2283e-7

5. Conclusion

The problem of Runge-Kutta Fehlberg method for solving fuzzy integro-differential equations has been studied. We have transformed the original problem to two parametric ODEs, which are solved by the Runge-Kutta Fehlberg method. Two numerical examples have been given. The obtain error rates from the Runge-Kutta Fehlberg method, the Runge-kutta method of order four, and the Euler method are summarized in Tables 1 and 2. It can be seen that the Runge-Kutta Fehlberg method is able to produce lower error rates as compared with those from the Runge-Kutta method of order four and the Euler method.

For further work, the solutions of higher-order fuzzy integro-differential equations will be investigated for solving a variety of problems.

Conflicts of Interest

The authors declare that there is no conflict of interest regarding the publication of this article.

References

- [1] R. C. Bassanezi, L. C. de Barros and P. A. Tonelli, Attractors and asymptotic stability for fuzzy dynamical systems, *Fuzzy Sets and Systems*, **113**, 473-483, (2000).
- [2] M. Guo, X. Xue and R. Li, Impulsive functional differential inclusions and fuzzy population models, *Fuzzy Sets and Systems*, **138**, 601-615, (2003).
- [3] V. Krivan and G. Colombo, A non-stochastic approach for modeling uncertainty in population dynamics, *Bulletin of Mathematical Biology*, **60**, 721-751, (1998).
- [4] M. Oberguggenberger and S. Pittschmann, Differential equations with fuzzy parameters, *Mathematical and Computer Modelling of Dynamical Systems*, **5**, 181-202, (1999).
- [5] M. Hanss, The transformation method for the simulation and analysis of systems with uncertain parameters, *Fuzzy Sets and Systems*, **130**, 277-289, (2002).
- [6] M. Hanss, *Applied fuzzy arithmetic*, Springer-Verlag Berlin Heidelberg, (2005).
- [7] R. C. Bassanezi and L. C. de Barros, A simple model of life expectancy with subjective parameters. *Kybernetes, The International Journal of Systems & Cybernetics*, **24**, 57-62, (1995).
- [8] S. S. L. Chang and L. A. Zadeh, On fuzzy mapping and control, *IEEE Trans. Systems Man Cybernet*, **2**, 30-34, (1972).
- [9] L.A. Zadeh, The concept of linguistic variable and its application to approximate reasoning, *Information Sciences*, **8**, 1, 99-249, (1975).
- [10] M. Mizumoto and K. Tanaka, The four operations of arithmetic on fuzzy numbers, *Syst. Comput. Controls*, **7**, 73-81, (1976).
- [11] D. Dubois and H. Prade, Operations on fuzzy numbers, *International Journal of systems science*, **9**, 613-626, (1978).
- [12] O. Abu Arqub, M. Al-Smadi and N. Shawagfeh, Solving Fredholm integro-differential equations using reproducing kernel Hilbert space method, *Applied Mathematics and Computation*, **219**, 8938-8948, (2013).
- [13] J. M. Cushing, *Integro differential equations and delay models in population dynamics*, Springer Science & Business Media, **20**, (2013).
- [14] J. M. Gushing, *Volterra integro differential equations in Population Dynamics*, in Mathematics of Biology, vol 80. Springer, Berlin, Heidelberg, 81-148, (2010).
- [15] S. Hajighasemi, T. Allahviranloo, M. Khezerloo, M. Khorasani and S. Salahshour, *Existence and uniqueness of solution of fuzzy voltra integro-differential equations*, in IPMU 2010, part II, CCIS 81. Springer, Berlin, 491-500, (2010).
- [16] K. Balachandran and K. Kanagarajan, Existence of solutions of fuzzy delay integro-differential equations with non local condition, *Journal of the Korean Society for Industrial and Applied Mathematics*, **9**, 65-74, (2005).
- [17] K. Balachandran and K. Kanagarajan, Existence of solutions of general nonlinear fuzzy Volterra-Fredholm integral equations, *Journal of Applied Mathematics and Stochastic Analysis*, **3**, 333-343, (2005).
- [18] P. Diamond, Theory and applications of fuzzy Volterra integral equations, *IEEE Transactions on Fuzzy Systems*, **10**, 97-102, (2002).
- [19] S. S. Behzadi, T. Allahviranloo and S. Abbasbandy, Solving fuzzy second-order nonlinear Volterra-Fredholm integro-differential equations by using Picard method, *Neural Computing and Applications*, **21**, 337-346, (2012).
- [20] K. Kanagarajan and R. Suresh, Runge-Kutta method for solving fuzzy differential equations under generalized differentiability, *Computational and Applied Mathematics*, **37**, 1294-1305, (2018).
- [21] K. Kanagarajan and R. Suresh, Numerical solution of fuzzy differential equations under generalized differentiability by Modified Euler method, *International Journal of Mathematical Engineering and Science*, **2**, 5-15, (2013).
- [22] B. Bede, A note on two-point boundary value problems associated with non-linear fuzzy differential equations, *Fuzzy Sets and Systems*, **157**, 986-989, (2006).
- [23] B. Bede, I. J. Rudas and A. L. Bencsik, First order linear fuzzy differential equations under generalized differentiability, *Information sciences*, **177**, 1648-1662, (2007).
- [24] J. J. Nieto, A. Khastan and K. Ivaz, Numerical solution of fuzzy differential equations under generalized differentiability, *Nonlinear Analysis: Hybrid Systems*, **3**, 700-707, (2009).
- [25] A. Ralston and P. Rabinowitz, *First course in numerical analysis*, Dover Publications, New York, (2001).
- [26] A. Seifi, T. Lotfi and T. Allahviranloo, A new efficient method using Fibonacci polynomials for solving of first-order fuzzy Fredholm-Volterra integro-differential equations, *Soft Computing*, **23**, 9777-9791, (2019).
- [27] S. Abbasbandy and M. S. Hashemi, A series solution of fuzzy integro-differential equations, *Journal of Fuzzy Set Valued Analysis*, **1**, 413-418, (2010).
- [28] S. Abbasbandy and M. S. Hashemi, Fuzzy integro-differential equations: formulation and solution using the variational iteration method, *Nonlinear Science Letters A*, **2012**, Article ID. jfsva-00066, (2010).

- [29] T. Allahviranloo, S. Abbasbandy, O. Sedaghatfar and P. Darabi, A new method for solving fuzzy integro-differential equation under generalized differentiability, *Neural Computing and Applications*, **21**, 191-196, (2012).
- [30] L. C. De Barros, R. C. Bassanezi and W. A. Lodwick, *First Course in Fuzzy Logic, Fuzzy Dynamical Systems, and Biomathematics*, Springer-Verlag Berlin An, (2016).
- [31] M. Friedman, M. Minga and A. Kandel, Fuzzy linear systems, *Fuzzy Sets and Systems*, **96**, 201-209, (1998).
- [32] A. A. Hemed, Formulation and solution of nth-order derivative fuzzy integro-differential equation using new iterative method with a reliable algorithm, *Journal of Applied Mathematics*, Article ID 325473, (2012).
- [33] O. Kaleva, Fuzzy differential equations, *Fuzzy Sets and Systems*, **24**, 301-317, (1987).
- [34] Wu Congxin and Ma Ming, On embedding problem of fuzzy number spaces, *Fuzzy Sets and Systems*, **44**, 33-38, (1991).
- [35] R. Goetschel and W. Vaxman, Elementary calculus, *Fuzzy Sets Systems*, **18**, 31-43, (1986).
- [36] M. L. Puri and D. Ralescu, Fuzzy random variables, *Journal of Mathematical Analysis and Applications*, **114**, 409-422, (1986).
- [37] M. Friedman, M. Ma and A. Kandel, Numerical solution of fuzzy differential and integral equations, *Fuzzy Sets and Systems*, **106**, 35-48, (1999).
- [38] B. Bede and S. G. Gal, Generalizations of the differentiability of fuzzy-number-valued functions with applications to fuzzy differential equation, *Fuzzy Sets and Systems*, **151**, 581-599, (2005).
- [39] Y. Chalco-Cano and H. Roman-Flores, On new solutions of fuzzy differential equations, *Chaos, Solitons & Fractals*, **38**, 112-119, (2008).
- [40] H. Hochstadt, *Integral equations*, John Willy & Sons, New York, (1973).
- [41] M. Mosleh and M. Otadi, Fuzzy Fredholm integro-differential equations with artificial neural networks, *Communications in Numerical Analysis*, 1-3, (2012).
- [42] P. Balasubramaniam and S. Muralisankar, Existence and uniqueness of fuzzy solution for the nonlinear fuzzy integro-differential equations, *Applied Mathematics Letters*, **14**, 455-462, (2001).



Porpattama

Hammachukiattikul received a Ph.D. degree in applied mathematics from the King Mongkut's University of Technology Thonburi (KMUTT), Bangkok, Thailand. She is currently working with the Department of Mathematics, Phuket Rajabhat University

(PKRU), Phuket, Thailand. Her research interests include mathematic model, climate change, atmospheric model, lyapunov theory and neural networks, stability analysis of dynamical systems, synchronization, and chaos theory.



Bundit Unyong received a Ph.D. degrees in Mathematics from Mahidol University in 2004, Thailand. He is currently working as an assistant professor in the Department of Mathematics, Faculty of Science and Technology, Phuket Rajabhat University, Phuket Thailand. He was experience and recipient

funding from Thai Government. His research interests include mathematic model, epidemic diseases control model under the effect of climate change, Oceanic model, atmospheric model, Lyapunov theory, and neural network.



R. Suresh received M.Sc., M.Phil., and Ph.D degrees in mathematics from Sri Ramakrishna Mission Vidyalaya College of Arts and Science, Bharathiar University, Coimbatore, Tamilnadu, India, in 2007, 2010 and 2017, respectively. Currently, he is working an assistant professor in the Department of Applied Mathematics, Sri Venkateswara

College of Engineering, Sriperumbudur, Tamilnadu, India. His current research interests include control theory and its applications for linear and nonlinear systems, time delay systems, neural networks and stochastic systems.



Grienggrai Rajchakit

was born in 1981. He received a B.S. (Mathematics), Thammasat University, Bangkok, Thailand, in 2003. He received a M.S. (Applied Mathematics), Chiangmai University, Chiangmai, Thailand, in 2005. He was awarded a Ph.D. (Applied Mathematics), King Mongkut's University of Technology Thonburi, Bangkok,

Thailand, in the field of mathematics with the specialized area of stability and control of neural networks. Currently, he is working as a lecturer at the Department of Mathematics, Faculty of Science, Maejo University, Chiangmai, Thailand. He was the recipient of Thailand Frontier Author Award by Thomson Reuters Web of Science in the year 2016 and TRF-OHEC-Scopus Researcher Awards by The Thailand Research Fund (TRF), Office of the Higher Education Commission (OHEC) and Scopus in the year 2016. His research interests are complex-valued NNs, complex dynamical networks, control theory, stability analysis, sampled-data control, multi-agent systems, T-S fuzzy theory, and cryptography, etc. Dr. Grienggrai Rajchakit serves as a reviewer for various SCI journals. He has authored and co-authored more than 111 research articles in various SCI journals.



R. Vadivel received B.Sc., M.Sc., and M.Phil. degrees in Mathematics from Sri Ramakrishna Mission Vidyalaya College of Arts and Science affiliated to Bharathiar University, Coimbatore, Tamil Nadu, India, in 2007, 2010, and 2012, respectively. He was awarded a Ph.D. degree in 2018

from the Department of Mathematics, Thiruvalluvar University, Vellore, Tamil Nadu, India. He was a Post-doctoral research fellow in the Research Center for Wind Energy Systems, Kunsan National University, Gunsan, South Korea, from 2018 to 2019. Currently, he is working as a lecturer in the Department of Mathematics, Faculty of Science and technology, Phuket Rajabhat university, Thailand.



Nallappan Gunasekaran received a BSC from Mahendra Arts and Science College, Periyar University, Salem, Tamil Nadu, India, in 2009. He received his post-graduation from Jamal Mohamed College affiliated to Bharathidasan University, Trichy, Tamil Nadu, India, in 2012. He was awarded a master of philosophy in 2013 from Bharathidasan University,

Trichy, India, and a Ph.D. degree from Thiruvalluvar University, Vellore, India, in 2018, in mathematics. He was a Post-doctoral research fellow in the Research Center for Wind Energy Systems, Kunsan National University, Gunsan, South Korea, from 2017 to 2018. Currently, he is working as a Post-doctoral research fellow in the Department of Mathematical Sciences, Shibaura Institute of Technology, Saitama, Japan.



Chee Peng Lim received a Ph.D. degree from the University of Sheffield, U.K., in 1997. He has authored and coauthored more than 450 technical papers in these areas. He is currently a professor with the Institute for Intelligent Systems Research and Innovation, Deakin University, Australia. His research interests

include computational intelligence-based systems for pattern recognition, data mining, fault detection, condition monitoring, decision support, and process optimization.