Applied Mathematical Sciences, Vol. 2, 2008, no. 1, 33 - 46

Numerical Solution of Fuzzy Integral Equations

M. Jahantigh ^{*a*}, **T. Allahviranloo** $b¹$ **and M. Otadi** ^{*c*}

^a Department of Mathematics, Islamic Azad University Zahedan Branch, Zahedan, Iran

 b Department of Mathematics, Tabarestan University, Chaloos, Iran</sup>

c Department of Mathematics, Islamic Azad University Firouzkooh Branch, Firouzkooh, Iran

Abstract

In this paper, a numerical procedure for solving fuzzy Fredholm integral equations of the second kind (*FIEs*) with arbitrary kernels have been investigated and residual minimization method is given and then the proposed algorithm is illustrated with solving some numerical examples.

Keywords: Fuzzy number, Fuzzy linear system, Fuzzy integral equations

1 Introduction

The concept of integration of fuzzy functions was first introduced by Dubois and Prade [5]. The topics of numerical methods for solving fuzzy integral equations have been rapidly growing in recent years and have been studies by authors of [6]. The numerical methods for fuzzy differential equations have been studied by S. Abbasbandy, T. Allahviranloo, [1, 2, 3] and others. Alternative approaches were later suggested by Goetschel and Vaxman [8], Kaleva [10] and others. The structure of this paper is organized as follows:

In section 2, some basic definitions and results on fuzzy numbers, fuzzy integral and the fuzzy linear system is brought. In Section 3, we propose a general method for solving fuzzy Fredholm integral equation of the second kind. In Section 4, we illustrate algorithm by solving some numerical examples. The conclusions are drawn in Section 5.

¹Corresponding author, e-mail: tofigh@allahviranloo.com

2 Preliminaries

Let us now introduce the notation needed in the rest of the paper. We will place a bar over a symbol if it represents a fuzzy number so \tilde{a}, b, \tilde{c} are all fuzzy numbers but a, b, c will denote real numbers. Parametric form of an arbitrary fuzzy number is given in [4] as follows. A fuzzy number \tilde{u} in parametric form is a pair (u, \overline{u}) of functions $u(r), \overline{u}(r), 0 \leq r \leq 1$, which satisfies the following requirements:

- 1. $u(r)$ is a bounded left continuous non-decreasing function over [0, 1],
- 2. $\overline{u}(r)$ is a bounded left continuous non-increasing function over [0, 1],
- 3. $u(r) < \overline{u}(r)$, $0 < r < 1$,

The set of all these fuzzy numbers is denoted by E. A crisp number α is simply represented by $u(r) = \overline{u}(r) = \alpha$, $0 \le r \le 1$. A popular fuzzy number is the triangular fuzzy number $\tilde{u} = (m, \alpha, \beta)$ which

$$
\tilde{u}(x) = \begin{cases}\n\frac{x-m}{\alpha} + 1, & m - \alpha \le x \le m, \\
\frac{m-x}{\beta} + 1, & m \le x \le m + \beta, \\
0, & otherwise.\n\end{cases}
$$

Its parametric form is

$$
\underline{u}(r) = m + \alpha(r - 1), \quad \overline{u}(r) = m + \beta(1 - r).
$$

By appropriate definitions the fuzzy number space $\{u(r), \overline{u}(r)\}\)$ becomes a convex cone E^1 which is then embedded isomorphically and isometrically into a Banach space.

Definition 1. The $n \times n$ dual linear system

$$
\begin{cases}\na_{11}\tilde{x}_1 + \cdots + a_{1n}\tilde{x}_n = \tilde{y}_1 + b_{11}\tilde{x}_1 + \cdots + b_{1n}\tilde{x}_n, \\
a_{21}\tilde{x}_1 + \cdots + a_{2n}\tilde{x}_n = \tilde{y}_2 + b_{21}\tilde{x}_1 + \cdots + b_{2n}\tilde{x}_n, \\
\vdots \qquad \vdots \qquad \vdots \qquad \vdots \\
a_{n1}\tilde{x}_1 + \cdots + a_{nn}\tilde{x}_n = \tilde{y}_n + b_{n1}\tilde{x}_1 + \cdots + b_{nn}\tilde{x}_n,\n\end{cases} (1)
$$

where the coefficient matrix $A = (a_{ij})$ and $B = (b_{ij}), 1 \le i, j \le n$ is a crisp $n \times n$ matrix, $\tilde{x}^t = (\tilde{x}_1, \ldots, \tilde{x}_n)$ be a $n \times 1$ vector of fuzzy numbers \tilde{x}_i and $\tilde{y}^t = (\tilde{y}_1, \ldots, \tilde{y}_n)$ be a $n \times 1$ vector of fuzzy numbers \tilde{y}_i is called a dual fuzzy linear system (DFLS).

For arbitrary fuzzy numbers $\tilde{x} = (x(r), \overline{x}(r)), \tilde{y} = (y(r), \overline{y}(r))$ and real number k , we may define the addition and the scalar multiplication of fuzzy numbers by using the extension principle as

- (a) $\tilde{x} = \tilde{y}$ if and only if $\underline{x}(r) = \underline{y}(r)$ and $\overline{x}(r) = \overline{y}(r)$,
- (b) $\tilde{x} + \tilde{y} = (\underline{x}(r) + \underline{y}(r), \overline{x}(r) + \overline{y}(r)),$ (c) $k\tilde{x} = \begin{cases} (k\underline{x}, k\overline{x}), & k \geq 0, \\ (k\underline{x}, k\underline{x}), & k > 0, \end{cases}$ $(k\overline{x}, k\underline{x}), \quad k < 0.$

Definition 2. A fuzzy number vector $(\tilde{x}_1, \tilde{x}_2, ..., \tilde{x}_n)^t$ given by

$$
\tilde{x}_i = (\underline{x}_i(r), \overline{x}_i(r)), \quad 1 \le i \le n, 0 \le r \le 1,
$$

is called a solution of the fuzzy linear system (1) if

$$
\begin{cases} \frac{\sum_{j=1}^{n} a_{ij} x_j}{\sum_{j=1}^{n} a_{ij} x_j} = \sum_{j=1}^{n} \frac{a_{ij} x_j}{\sum_{j=1}^{n} a_{ij} x_j} = \sum_{j=1}^{n} \frac{b_{ij} x_j}{\sum_{j=1}^{n} a_{ij} x_j} = \sum_{j=1}^{n} \overline{a_{ij} x_j} = \overline{y}_i = \overline{\sum_{j=1}^{n} b_{ij} x_j} = \sum_{j=1}^{n} \overline{b}_{ij} x_j. \end{cases}
$$

If, for a particular $i, a_{ij} > 0$ and $b_{ij} > 0, 1 \le j \le n$, we simply get

$$
\sum_{j=1}^n a_{ij} \underline{x}_j = \underline{y}_i + \sum_{j=1}^n b_{ij} \underline{x}_j, \qquad \sum_{j=1}^n a_{ij} \overline{x}_j = \overline{y}_i + \sum_{j=1}^n b_{ij} \overline{x}_j.
$$

The following theorem guarantees the existence of a fuzzy solution for general case. Consider the dual fuzzy linear system (6), and transform its $n \times n$ coefficient matrix A and B in to $(2n) \times (2n)$ matrices as:

$$
\begin{cases}\ns_{11}\underline{x}_{1} + \cdots + s_{1n}\underline{x}_{n} + s_{1,n+1}(-\overline{x}_{1}) + \cdots + s_{1,2n}(-\overline{x}_{n}) = \underline{y}_{1} + \\
& t_{11}\underline{x}_{1} + \cdots + t_{1n}\underline{x}_{n} + t_{1,n+1}(-\overline{x}_{1}) + \cdots + t_{1,2n}(-\overline{x}_{n}), \\
\vdots \\
s_{n1}\underline{x}_{1} + \cdots + s_{nn}\underline{x}_{n} + s_{n,n+1}(-\overline{x}_{1}) + \cdots + s_{n,2n}(-\overline{x}_{n}) = \underline{y}_{n} + \\
& t_{n1}\underline{x}_{1} + \cdots + t_{nn}\underline{x}_{n} + t_{n,n+1}(-\overline{x}_{1}) + \cdots + t_{n,2n}(-\overline{x}_{n}), \\
s_{n+1,1}\underline{x}_{1} + \cdots + s_{n+1,n}\underline{x}_{n} + s_{n+1,n+1}(-\overline{x}_{1}) + \cdots + s_{n+1,2n}(-\overline{x}_{n}) = -\overline{y}_{1} + \\
& t_{n+1,1}\underline{x}_{1} + \cdots + t_{n+1,n}\underline{x}_{n} + t_{n+1,n+1}(-\overline{x}_{1}) + \cdots + t_{n+1,2n}(-\overline{x}_{n}), \\
\vdots \\
s_{2n,1}\underline{x}_{1} + \cdots + s_{2n,n}\underline{x}_{n} + s_{2n,n+1}(-\overline{x}_{1}) + \cdots + s_{2n,2n}(-\overline{x}_{n}) = -\overline{y}_{n} + \\
& t_{2n,1}\underline{x}_{1} + \cdots + t_{2n,n}\underline{x}_{n} + t_{2n,n+1}(-\overline{x}_{1}) + \cdots + t_{2n,2n}(-\overline{x}_{n}),\n\end{cases}
$$

where s_{ij} and t_{ij} are determined as follows:

$$
a_{ij} \ge 0 \Longrightarrow s_{ij} = a_{ij}, \ s_{i+n,j+n} = a_{ij},
$$

\n
$$
a_{ij} < 0 \Longrightarrow s_{i,j+n} = -a_{ij}, \ s_{i+n,j} = -a_{ij},
$$

\n
$$
b_{ij} \ge 0 \Longrightarrow t_{ij} = b_{ij}, \ t_{i+n,j+n} = b_{ij},
$$

\n
$$
b_{ij} < 0 \Longrightarrow t_{i,j+n} = -b_{ij}, \ t_{i+n,j} = -b_{ij},
$$
\n
$$
(2)
$$

and any s_{ij} and t_{ij} which is not determined by (2) is zero. Using matrix notation we get

$$
SX = Y + TX,\tag{3}
$$

therefore, we have:

$$
(S-T)X = Y,\t\t(4)
$$

where $S = (s_{ij}) \ge 0$ and $T = (t_{ij}) \ge 0, 1 \le i, j \le 2n$, and

$$
X = \begin{bmatrix} \frac{x_1}{\vdots} \\ \frac{x_n}{\ddots} \\ \frac{x_n}{\ddots} \\ \frac{x_n}{\ddots} \end{bmatrix}, \quad Y = \begin{bmatrix} \frac{y_1}{\vdots} \\ \frac{y_n}{\ddots} \\ \frac{y_n}{\ddots} \\ \frac{x_n}{\ddots} \end{bmatrix}.
$$
 (5)

For example:

Consider the dual fuzzy linear system

$$
\begin{cases} \tilde{x}_1 - \tilde{x}_2 = \tilde{y}_1 + 2\tilde{x}_1 + \tilde{x}_2, \\ \tilde{x}_1 + 2\tilde{x}_2 = \tilde{y}_2 + \tilde{x}_1 - 2\tilde{x}_2. \end{cases} \tag{6}
$$

Let $\underline{y}_1 = r, \overline{y}_1 = 2 - r$ and $\underline{y}_2 = 4 + r, \overline{y}_2 = 7 - 2r$, the extended 4×4 matrices are

$$
S = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 1 & 2 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix}, \quad T = \begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 0 & 0 & 2 \\ 0 & 0 & 2 & 1 \\ 0 & 2 & 1 & 0 \end{bmatrix},
$$

and

$$
Y = \begin{bmatrix} r \\ 4+r \\ r-2 \\ 2r-7 \end{bmatrix}, \quad X = \begin{bmatrix} \frac{x_1}{x_2} \\ \frac{x_2}{-x_1} \\ -\overline{x}_2 \end{bmatrix}.
$$

We obtain that the system (6) is equivalent to the function equation system

$$
SX = Y + TY.
$$

Consequently,

$$
\begin{bmatrix} 1 & 0 & 0 & 1 \ 1 & 2 & 0 & 0 \ 0 & 1 & 1 & 0 \ 0 & 0 & 1 & 2 \ \end{bmatrix} \begin{bmatrix} \frac{x_1}{x_2} \\ \frac{x_2}{-x_1} \\ \frac{-x_1}{-x_2} \end{bmatrix} = \begin{bmatrix} r \\ 4+r \\ r-2 \\ 2r-7 \end{bmatrix} + \begin{bmatrix} 2 & 1 & 0 & 0 \ 1 & 0 & 0 & 2 \ 0 & 0 & 2 & 1 \ 0 & 2 & 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{x_1}{x_2} \\ \frac{x_2}{-x_1} \\ \frac{-x_1}{-x_2} \end{bmatrix}.
$$

Also,

$$
(S-T)X = Y.
$$

The structure of S and T implies that:

$$
S = \left(\begin{array}{cc} C & D \\ D & C \end{array} \right), \quad T = \left(\begin{array}{cc} E & F \\ F & E \end{array} \right),
$$

where C and E contains the positive entries of A and B respectively, and D and F the absolute values of the negative entries of A and B, i.e. $A = C - D$ and $B = E - F$. Therefore

$$
S-T=\left(\begin{array}{cc} C-E & D-F \\ D-F & C-E \end{array}\right),
$$

Theorem 1. The matrix $S - T$ is nonsingular if and if the matrix $(C + D)$ – $(E + F)$ and $(C + F) - (E + D)$ are both nonsingular.

Proof. The same as the proof of Theorem 1 in [7].

Assuming that $S - T$ is nonsingular we obtain of the Eq. (4)

$$
X = (S - T)^{-1}Y\tag{7}
$$

Theorem 2. If $(S - T)^{-1}$ exists it must have the same structure as S, i.e.

$$
(S-T)^{-1} = \left(\begin{array}{cc} G & H \\ H & G \end{array}\right),
$$

and

$$
G = \frac{1}{2} [((C + D) - (E + F))^{-1} + ((C + F) - (E + D))^{-1}],
$$

\n
$$
H = \frac{1}{2} [((C + D) - (E + F))^{-1} - ((C + F) - (E + D))^{-1}],
$$

Proof. see [7].

Theorem 3. The unique solution X of Eq. (7) is a fuzzy vector for arbitrary Y if and only if $(S-T)^{-1}$ is nonnegative, i.e.

$$
((S-T)^{-1})_{ij} \ge 0
$$
, $1 \le i \le 2n$, $1 \le j \le 2n$.

Proof. The same as the proof of Theorem 3 in [7].

Definition 3. Let $X = \{(\underline{x}_i(r), \overline{x}_i(r)), 1 \leq i \leq n\}$ denotes the unique solution of (3), if $y_i(r), \overline{y}_i(r)$ are linear functions of r, then the fuzzy number vector $U = \{(\underline{u}_i(\overline{r}), \overline{u}_i(r)), 1 \leq i \leq n\}$ defined by

$$
\underline{u}_i(r) = \min{\underline{x}_i(r), \overline{x}_i(r), \underline{x}_i(1)},
$$

$$
\overline{u}_i(r) = \max{\underline{x}_i(r), \overline{x}_i(r), \underline{x}_i(1)}.
$$

is called the fuzzy solution of (3). If $(\underline{x}_i(r), \overline{x}_i(r))$, $1 \leq i \leq n$, are all fuzzy numbers then $\underline{u}_i(r) = \underline{x}_i(r), \overline{u}_i(r) = \overline{x}_i(r)$, and then U is called a strong fuzzy solution. Otherwise, U is a weak fuzzy solution.

We will next define a metric D in E^1 and the fuzzy function notation [8]. **Definition 4.** For arbitrary fuzzy numbers $\tilde{u} = (u, \overline{u})$ and $\tilde{v} = (v, \overline{v})$ the quantity

$$
D(\tilde{u}, \tilde{v}) = \sup_{0 \le r \le 1} \{ \max[\vert \underline{u}(r) - \underline{v}(r) \vert, \vert \overline{u}(r) - \overline{v}(r) \vert] \} \tag{8}
$$

is the distance between \tilde{u} and \tilde{v} . It is shown [11] that (E^1, D) is a complete metric space.

Definition 5. Let $\tilde{U} = (\tilde{u_1}, \ldots, \tilde{u_n})$ and $\tilde{V} = (\tilde{v_1}, \ldots, \tilde{v_n}), \dot{D}(\tilde{U}, \tilde{V})$ is:

$$
\dot{D}(\tilde{U}, \tilde{V}) = \begin{bmatrix} D(\tilde{u}_1, \tilde{v}_1) \\ \vdots \\ D(\tilde{u}_n, \tilde{v}_n) \end{bmatrix}
$$
\n(9)

Definition 6. A function $\tilde{f}: \mathbb{R}^1 \longrightarrow E^1$ is called a fuzzy function. If for arbitrary fixed $t_0 \in \mathbb{R}^1$ and $\varepsilon > 0$, a $\delta > 0$ such that

$$
|t - t_0| < \delta \Longrightarrow D[\tilde{f}(t), \tilde{f}(t_0)] < \epsilon \tag{10}
$$

exist, f is said to be continuous.

3 Fuzzy integral equations

The Fredholm integral equation of the second kind is

$$
x(s) = f(s) + \lambda \int_{a}^{b} k(s, t)x(t)dt
$$
 (11)

where $\lambda > 0$, $k(s,t)$ is an arbitrary kernel function over the square $a \leq s, t \leq b$ and $f(t)$ is a function of $t : a \le t \le b$, [9]. If $f(t)$ is a crisp function then the solution of Eqs. (11) is crisp as well. However, if $f(t)$ is a fuzzy function this equation may only possess fuzzy solution. Therefore, we have

$$
\tilde{x}(s) = \tilde{f}(s) + \lambda \int_{a}^{b} k(s, t)\tilde{x}(t)dt.
$$
\n(12)

Sufficient conditions for the existence of a unique solution to the fuzzy Fredholm integral equation of the second kind, i.e. to Eq. (12) where $f(t)$ is a fuzzy function, are given in [12].

We consider now the numerical solution of fuzzy Fredholm integral equations of the second kind Eq. (12), which we write in the form:

$$
\tilde{x} = \tilde{f} + \lambda K \tilde{x}.\tag{13}
$$

The exact solution of integral equation Eq. (12) is:

$$
\tilde{x}(s) = \sum_{i=1}^{\infty} \tilde{a}_i h_i(s) \tag{14}
$$

in truncated form

$$
\tilde{x}(s) \approx \tilde{x}_n(s) = \sum_{i=1}^n \tilde{a}_i h_i(s),\tag{15}
$$

where the set $\{h_i\}$ is complete and orthogonal in $\ell^2(a, b)$ (see [?]). For finding approximation solution we must indicate coefficients \tilde{a}_i .

From Eq. (15) we obtain

$$
\sum_{j=1}^{n} \tilde{a}_{j} h_{j}(s) = \tilde{f}(s) + \lambda \sum_{j=1}^{n} \tilde{a}_{j} \int_{a}^{b} k(s, t) h_{j}(t) dt.
$$
 (16)

We have *n* unknown parameters in the form $\tilde{a}_1, \tilde{a}_2, \ldots, \tilde{a}_n$ which for finding them, we need to *n* equation, so by using *n* point s_1, s_2, \ldots, s_n in interval [a, b]:

$$
\sum_{j=1}^{n} h_j(s_i)\tilde{a}_j = \tilde{f}(s_i) + \lambda \sum_{j=1}^{n} \int_a^b k(s_i, t)h_j(t)dt \tilde{a}_j \qquad i = 1, \cdots, n \quad (17)
$$

therefore we have:

$$
A\tilde{a} = \tilde{f} + B\tilde{a},\tag{18}
$$

where the coefficients matrix $A = (a_{ij}), 1 \le i, j \le n$, and $B = (b_{ij}), 1 \le i, j \le n$ n, are crisp and $\tilde{f} = (\tilde{f}_i), 1 \leq i \leq n$, is an arbitrary fuzzy number vector, where

$$
a_{ij} = h_j(s_i), \qquad b_{ij} = \lambda \int_a^b k(s_i, t) h_j(t) dt, \qquad i, j = 1, \cdots, n
$$

3.1 Residual minimization method

The simplest method conceptually again appeals to approximation theory. We write the integral equation in the form (again we set $\lambda = 1$)

$$
L\tilde{x} = \tilde{f}, \quad L = I - K,\tag{19}
$$

and introduce the residual function r_n and error function ε_n

$$
r_n = \dot{D}(\tilde{f}, L\tilde{x}_n),\tag{20}
$$

$$
\varepsilon_n = \dot{D}(\tilde{x}, \tilde{x}_n). \tag{21}
$$

To compute r_n requires no knowledge of \tilde{x} but, since $\dot{D}(\tilde{f},L\tilde{x})=0$, we have the identity:

$$
r_n = \dot{D}(\tilde{f}, L\tilde{x}_n) - \dot{D}(\tilde{f}, L\tilde{x}) = L\dot{D}(\tilde{x}_n, \tilde{x}) = L\varepsilon_n.
$$
 (22)

From (19) and (22) we have at once

$$
\parallel r_n \parallel \leq (1 + \parallel K \parallel) \parallel \varepsilon_n \parallel.
$$
\n(23)

That is,

$$
\parallel \varepsilon_n \parallel \ge \frac{\parallel r_n \parallel}{1 + \parallel K \parallel}. \tag{24}
$$

Thus a small residual is a necessary condition for a small error. We would rather have an upper bound on ε_n of course; this is harder to provide in general and we content ourselves for now with the following. We rewrite (22) as

$$
\varepsilon_n = r_n + K \varepsilon_n
$$

whence

$$
\|\varepsilon_n\|\leq \|r_n\| + \|K\| \cdot \|\varepsilon_n\| \quad \text{and hence if } \|K\| < 1 \tag{25}
$$
\n
$$
\|\varepsilon_n\| \leq \frac{\|r_n\|}{1 - \|K\|}.\tag{26}
$$

$$
\parallel \varepsilon_n \parallel \leq \frac{\parallel r_n \parallel}{1 - \parallel K \parallel}. \tag{26}
$$

4 Numerical examples

Example 1. Consider the fuzzy integral equation (12) where

$$
\frac{f(s;r) = s^3(r^2 + r)}{\overline{f}(s;r) = s^3(4 - r^3 - r)},
$$

and kernel

$$
k(s,t) = s + 1,
$$
 $-1 \le s, t \le 1.$

and $a = -1, b = 1$. The exact solution in this case is given by

$$
\underline{x}(s;r) = s^3(r^2+r),
$$

$$
\overline{x}(s;r) = s^3(4-r^3-r).
$$

Let

$$
h_1(s) = 1, h_2(s) = s^3,
$$

and

$$
s_1 = -1, s_2 = 1.
$$

From the Eqs. (17) and (18) :

$$
\begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \end{bmatrix} = \begin{bmatrix} \tilde{f}(s_1) \\ \tilde{f}(s_2) \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 4 & 0 \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \end{bmatrix},
$$

the Extended 4×4 matrices are

$$
S = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}, \quad T = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 4 & 0 \end{bmatrix}
$$

and the solution of Eq. (7) is:

$$
\begin{bmatrix}\n\underline{a}_1 \\
\underline{a}_2 \\
-\overline{a}_1 \\
-\overline{a}_2\n\end{bmatrix} = (S - T)^{-1}F = \begin{bmatrix}\n-\frac{r^3}{4} - \frac{r^2}{4} - \frac{r}{2} + 1 \\
-\frac{3}{4}r^3 + \frac{r^2}{4} - \frac{r}{2} + 3 \\
-\frac{r^3}{4} - \frac{r^2}{4} - \frac{r}{2} + 1 \\
\frac{r^3}{4} - \frac{3}{4}r^2 - \frac{r}{2} - 1\n\end{bmatrix}.
$$
\n(27)

The fact that \tilde{a}_1 and \tilde{a}_2 are not fuzzy numbers. Therefore the fuzzy solution of the Eq. (27) is a weak fuzzy solution. The fuzzy approximate solution in this case is given by

$$
\tilde{x}(s) \approx \tilde{x}_2(s) = \sum_{i=1}^2 \tilde{a}_i h_i(s)
$$

where

$$
\begin{array}{l} \tilde{a}_1=(\frac{r^3}{4}+\frac{r^2}{4}+\frac{r}{2}-1,-\frac{r^3}{4}-\frac{r^2}{4}-\frac{r}{2}+1) \\ \tilde{a}_2=(-\frac{r^3}{4}+\frac{3}{4}r^2+\frac{r}{2}+1,-\frac{3}{4}r^3+\frac{r^2}{4}-\frac{r}{2}+3).\end{array}
$$

Example 2. Consider the fuzzy integral equation (12) where

$$
\underline{f}(s;r) = (\frac{s}{2} - \frac{1}{3})r,
$$

$$
\overline{f}(s;r) = (\frac{s}{2} - \frac{1}{3})(2-r),
$$

and kernel

$$
k(s,t) = s+t, \qquad 0 \le s, t \le 1.
$$

and $a = 0, b = 1$. The exact solution in this case is given by

$$
\underline{x}(s;r) = sr,
$$

$$
\overline{x}(s;r) = s(2-r).
$$

Let

$$
h_1(s) = s, h_2(s) = s^3, h_3(s) = s^5,
$$

and

$$
s_1 = 0, s_2 = \frac{1}{2}, s_3 = 1.
$$

From the Eqs. (17) and (18) :

$$
\begin{bmatrix} 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{8} & \frac{1}{32} \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \\ \tilde{a}_3 \end{bmatrix} = \begin{bmatrix} \tilde{f}(s_1) \\ \tilde{f}(s_2) \\ \tilde{f}(s_3) \end{bmatrix} + \begin{bmatrix} \frac{1}{3} & \frac{1}{5} & \frac{1}{7} \\ \frac{7}{12} & \frac{13}{40} & \frac{19}{84} \\ \frac{5}{6} & \frac{9}{20} & \frac{13}{42} \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \\ \tilde{a}_3 \end{bmatrix},
$$

the Extended 6×6 matrices are

$$
S = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{8} & \frac{1}{32} & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{8} & \frac{1}{32} \\ 0 & 0 & 0 & 1 & 1 & 1 \end{bmatrix}, \quad T = \begin{bmatrix} \frac{1}{3} & \frac{1}{5} & \frac{1}{7} & 0 & 0 & 0 \\ \frac{7}{12} & \frac{13}{40} & \frac{19}{84} & 0 & 0 & 0 \\ \frac{5}{6} & \frac{9}{20} & \frac{13}{42} & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{3} & \frac{1}{5} & \frac{1}{7} \\ 0 & 0 & 0 & \frac{7}{6} & \frac{13}{20} & \frac{19}{42} \\ 0 & 0 & 0 & \frac{5}{6} & \frac{9}{20} & \frac{13}{42} \end{bmatrix}
$$

and the solution of Eq. (7) is:

 \lceil

 $\overline{}$ \vert \vert \vert \vert \vert $\overline{}$

$$
\begin{bmatrix}\n\underline{a}_1 \\
\underline{a}_2 \\
\underline{a}_3 \\
-\overline{a}_1 \\
-\overline{a}_2 \\
-\overline{a}_3\n\end{bmatrix} = (S - T)^{-1}F = \begin{bmatrix}\nr \\
0 \\
0 \\
-2+r \\
0 \\
0\n\end{bmatrix},
$$
\n(28)

i.e.

$$
\begin{aligned}\n\tilde{a}_1 &= (r, 2 - r) \\
\tilde{a}_2 &= (0, 0) \\
\tilde{a}_3 &= (0, 0).\n\end{aligned}
$$

Here $\underline{a}_1 \le \overline{a}_1, \underline{a}_2 \le \overline{a}_2, \underline{a}_3 \le \overline{a}_3$ are monotonic decreasing functions. Therefore the fuzzy solution of the Eq. (28) is a strong fuzzy solution. The fuzzy approximate solution in this case is a approximate solution given by

$$
\tilde{x}(s) \approx \tilde{x}_3(s) = \sum_{i=1}^3 \tilde{a}_i h_i(s)
$$

Example 3. Consider the fuzzy integral equation (12) where

$$
\frac{f(s;r) = -\frac{2}{\pi}\cos s(r^2 + r)},
$$

$$
\overline{f}(s;r) = -\frac{2}{\pi}\cos s(3 - r),
$$

and kernel

$$
k(s,t) = \cos(s-t), \qquad \lambda = \frac{4}{\pi},
$$

and $a = 0, b = \frac{\pi}{2}$. The exact solution in this case is given by

$$
\underline{x}(s;r) = \sin s(r^2 + r) = (s - \frac{s^3}{3!} + \frac{s^5}{5!} - \dots)(r^2 + r),
$$

$$
\overline{x}(s;r) = \sin s(3 - r) = (s - \frac{s^3}{3!} + \frac{s^5}{5!} - \dots)(3 - r).
$$

Let

$$
h_1(s) = s, h_2(s) = s^3, h_3(s) = s^5,
$$

and

$$
s_1 = 0, s_2 = \frac{\pi}{4}, s_3 = \frac{\pi}{2} - 0.01.
$$

From the Eqs. (17) and (18) :

$$
\begin{bmatrix}\n0 & 0 & 0 \\
\frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 \\
\frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5\n\end{bmatrix}\n\begin{bmatrix}\n\tilde{a}_1 \\
\tilde{a}_2 \\
\tilde{a}_3\n\end{bmatrix} =\n\begin{bmatrix}\n\tilde{f}(s_1) \\
\tilde{f}(s_2) \\
\tilde{f}(s_3)\n\end{bmatrix} +\n\begin{bmatrix}\n0.726765132 & 0.574239947 & 0.6913436 \\
1.4142126 & 1.668453 & 2.6463 \\
1.2805 & 1.791 & 3.0578\n\end{bmatrix}\n\begin{bmatrix}\n\tilde{a}_1 \\
\tilde{a}_2 \\
\tilde{a}_3\n\end{bmatrix},
$$

the Extended 6×6 matrices are

$$
S = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 & 0 & 0 & 0 & 0 \\ \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 \\ 0 & 0 & 0 & \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \\ 0 & 0 & 0 & \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \end{bmatrix},
$$

\n
$$
T = \begin{bmatrix} 0.726765132 & 0.574239947 & 0.6913436 & 0 & 0 & 0 \\ 1.4142126 & 1.668453 & 2.6463 & 0 & 0 & 0 \\ 1.2805 & 1.791 & 3.0578 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.726765132 & 0.574239947 & 0.6913436 \\ 0 & 0 & 0 & 1.4142126 & 1.668453 & 2.6463 \\ 0 & 0 & 0 & 1.2805 & 1.791 & 3.0578 \end{bmatrix}
$$

and the solution of Eq. (7) is:

$$
\begin{bmatrix}\n\underline{a}_{1} \\
\underline{a}_{2} \\
\underline{a}_{3} \\
-\overline{a}_{1} \\
-\overline{a}_{2} \\
-\overline{a}_{3}\n\end{bmatrix} = (S - T)^{-1}F = \begin{bmatrix}\n1.0112r^{2} + 1.012r \\
-0.196r^{2} - 0.196r \\
0.0192r^{2} + 0.0192r \\
1.012r - 3.036 \\
0.587 - 0.196r \\
0.00481r - 0.057\n\end{bmatrix}.
$$
\n(29)

i.e.

$$
\begin{aligned}\n\tilde{a}_1 &= (1.0112r^2 + 1.012r, -1.012r + 3.036) \\
\tilde{a}_2 &= (-0.196r^2 - 0.196r, -0.587 + 0.196r) \\
\tilde{a}_3 &= (0.0192r^2 + 0.0192r, -0.00481r + 0.057).\n\end{aligned}
$$

The fact that \tilde{a}_2 is not a fuzzy number. Therefore the fuzzy solution of the Eq. (29) is a weak fuzzy solution. The fuzzy approximate solution in this case is a approximate solution given by

$$
\tilde{x}(s) \approx \tilde{x}_3(s) = \sum_{i=1}^3 \tilde{a}_i h_i(s)
$$

where

$$
\begin{aligned}\n\tilde{a}_1 &= (1.0112r^2 + 1.012r, -1.012r + 3.036) \\
\tilde{a}_2 &= (-0.587 + 0.196r, -0.196r^2 - 0.196r) \\
\tilde{a}_3 &= (0.0192r^2 + 0.0192r, -0.00481r + 0.057).\n\end{aligned}
$$

5 Conclusions

In this paper, we proposed a numerical method for solving fuzzy Fredholm integral equation of the second kind. The resulted approximate solutions from expansion method may be a strong or weak fuzzy solutions.

References

- [1] S. Abbasbandy, T. Allahviranloo, Oscar Lopez-Pouso, Juan J. Nieto, Numerical Methods for Fuzzy Differential Inclusions, Journal of Computer and Mathematics With Applications 48 (2004) 1633-1641.
- [2] S. Abbasbandy and T. Allahviranloo, Numerical solution of fuzzy differential equations, Mathematical and computational Applications, 7 (2002), No.1, 41-52.
- [3] S. Abbasbandy and T. Allahviranloo, Numerical solution of fuzzy differential equations, Numerical Solutions of Fuzzy Differential Equations By Taylor Method, Computational Methods in Applied Mathematics, 2 (2002), No.2, 113-124.
- [4] W. Cong-Xin and M. Ming, Embeddingproblem of fuzzy number space: Part I, Fuzzy Sets and Systems 44 (1991) 33-38.
- [5] D. Dubois, H. Prade, Towards fuzzy differential calculus, Fuzzy Sets and System 8 (1982) 1-7, 105-116,225-233.
- [6] M. Friedman, M. Ma, A. Kandel, Numerical Solutions of fuzzy differential equations and integral equations, Fuzzy Sets and Systems 106 (1999) 35- 48.
- [7] M. Friedman, M. Ming and A. Kandel, Duality in Fuzzy linear systems, Fuzzy Sets and Systems 109 (2000) 55-58.
- [8] R. Goetschel, W. Voxman, Elemenetary calculus, Fuzzy Sets and Systems 18 (1986) 31-43.
- [9] H. Hochstadt, Integral Equations, Wiley, New York, 1973, pp. 1-24.
- [10] O. Kaleva, Fuzzy differential equations, Fuzzy Sets and Systems 24 (1987) 301-317.
- [11] M.L. Puri, D. Ralescu, Fuzzy random variables, J. Math. Anal. Appl. 114 (1986) 409-422.
- [12] Wu Congxin, Ma Ming, On the integrals,series and integral equations of fuzzy set-valued functions, J. Harbin Inst. Technol. 21 (1990) 11-19.

Received: May 29, 2007