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

Numerical Solution of Fuzzy Integral Equations

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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.

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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 $\tilde{a}, \tilde{b}, \tilde{c}$ 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 $(\underline{u}, \overline{u})$ of functions $\underline{u}(r), \overline{u}(r), 0 \leq r \leq 1$, which satisfies the following requirements:

- 1. $\underline{u}(r)$ 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. $\underline{u}(r) \leq \overline{u}(r), 0 \leq r \leq 1$,

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

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

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 $\{\underline{u}(r), \overline{u}(r)\}$ 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

$$\begin{cases} a_{11}\tilde{x}_{1} + \cdots + a_{1n}\tilde{x}_{n} = \tilde{y}_{1} + b_{11}\tilde{x}_{1} + \cdots + b_{1n}\tilde{x}_{n}, \\ a_{21}\tilde{x}_{1} + \cdots + a_{2n}\tilde{x}_{n} = \tilde{y}_{2} + b_{21}\tilde{x}_{1} + \cdots + b_{2n}\tilde{x}_{n}, \\ \vdots & \vdots & \vdots \\ a_{n1}\tilde{x}_{1} + \cdots + a_{nn}\tilde{x}_{n} = \tilde{y}_{n} + b_{n1}\tilde{x}_{1} + \cdots + b_{nn}\tilde{x}_{n}, \end{cases}$$
(1)

where the coefficient matrix $A = (a_{ij})$ and $B = (b_{ij}), 1 \leq i, j \leq n$ 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}_j 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 $\tilde{x} = (\underline{x}(r), \overline{x}(r)), \tilde{y} = (\underline{y}(r), \overline{y}(r))$ 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) $k\tilde{x} = \begin{cases} (k\underline{x}, k\overline{x}), & k \ge 0, \\ (k\overline{x}, k\underline{x}), & k < 0. \end{cases}$

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

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

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:

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

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

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

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

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

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

(2)

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

$$SX = Y + TX, (3)$$

therefore, we have:

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

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

$$X = \begin{bmatrix} \frac{\underline{x}_1}{\vdots} \\ \frac{\underline{x}_n}{-\overline{x}_1} \\ \vdots \\ -\overline{x}_n \end{bmatrix}, \quad Y = \begin{bmatrix} \frac{\underline{y}_1}{\vdots} \\ \frac{\underline{y}_n}{-\overline{y}_1} \\ \vdots \\ -\overline{y}_n \end{bmatrix}.$$
(5)

For example:

Consider the dual fuzzy linear system

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

$$Y = \begin{bmatrix} r \\ 4+r \\ r-2 \\ 2r-7 \end{bmatrix}, \quad X = \begin{bmatrix} \underline{x}_1 \\ \underline{x}_2 \\ -\overline{x}_1 \\ -\overline{x}_2 \end{bmatrix}.$$

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

$$SX = Y + TY.$$

Consequently,

$$\begin{bmatrix} 1 & 0 & 0 & 1 \\ 1 & 2 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} \underline{x}_1 \\ \underline{x}_2 \\ -\overline{x}_1 \\ -\overline{x}_2 \end{bmatrix} = \begin{bmatrix} r \\ 4+r \\ r-2 \\ 2r-7 \end{bmatrix} + \begin{bmatrix} 2 & 1 & 0 & 0 \\ 1 & 0 & 0 & 2 \\ 0 & 0 & 2 & 1 \\ 0 & 2 & 1 & 0 \end{bmatrix} \begin{bmatrix} \underline{x}_1 \\ \underline{x}_2 \\ -\overline{x}_1 \\ -\overline{x}_2 \end{bmatrix}.$$

Also,

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

The structure of S and T implies that:

$$S = \begin{pmatrix} C & D \\ D & C \end{pmatrix}, \quad T = \begin{pmatrix} E & F \\ F & E \end{pmatrix},$$

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 - Dand 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)

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

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

$$(S-T)^{-1} = \begin{pmatrix} G & H \\ H & G \end{pmatrix},$$

and

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

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

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.

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

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 $\underline{y}_i(r), \overline{y}_i(r)$ are linear functions of r, then the fuzzy number vector $U = \{(\underline{u}_i(r), \overline{u}_i(r)), 1 \leq i \leq n\}$ 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 $\tilde{u} = (\underline{u}, \overline{u})$ and $\tilde{v} = (\underline{v}, \overline{v})$ the quantity

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

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

$$|t - t_0| < \delta \Longrightarrow D[\tilde{f}(t), \tilde{f}(t_0)] < \epsilon$$
(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 $t : a \leq t \leq b$, [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

$$\tilde{x}(s) = \tilde{f}(s) + \lambda \int_{a}^{b} k(s,t)\tilde{x}(t)dt.$$
(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 $\tilde{f}(t)$ 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 $A = (a_{ij}), 1 \leq i, j \leq n$, and $B = (b_{ij}), 1 \leq i, j \leq n$, are crisp and $\tilde{f} = (\tilde{f}_i), 1 \leq i \leq n$, is an arbitrary fuzzy number vector, where

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

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

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

That is,

$$\|\varepsilon_n\| \ge \frac{\|r_n\|}{1+\|K\|}.$$
(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

$$\|\varepsilon_n\| \le \|r_n\| + \|K\| \cdot \|\varepsilon_n\| \quad and hence if \|K\| < 1$$

$$\|\varepsilon_n\| = \|\varepsilon_n\|$$

$$\|\varepsilon_n\| \le \|\varepsilon_n\|$$

$$\|\varepsilon_n\| \le \|\varepsilon_n\|$$

$$\|\varepsilon_n\| \le \|\varepsilon_n\|$$

$$(25)$$

$$\|\varepsilon_n\| \le \frac{\|T_n\|}{1-\|K\|}.$$
 (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

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

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

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

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:

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

$$\tilde{\tilde{a}}_1 = \left(\frac{r^3}{4} + \frac{r^2}{4} + \frac{r}{2} - 1, -\frac{r^3}{4} - \frac{r^2}{4} - \frac{r}{2} + 1\right)$$
$$\tilde{\tilde{a}}_2 = \left(-\frac{r^3}{4} + \frac{3}{4}r^2 + \frac{r}{2} + 1, -\frac{3}{4}r^3 + \frac{r^2}{4} - \frac{r}{2} + 3\right).$$

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

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

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

and the solution of Eq. (7) is:

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

i.e.

$$\tilde{a}_1 = (r, 2 - r)$$

 $\tilde{a}_2 = (0, 0)$

 $\tilde{a}_3 = (0, 0).$

Here $\underline{a}_1 \leq \overline{a}_1, \underline{a}_2 \leq \overline{a}_2, \underline{a}_3 \leq \overline{a}_3$ 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

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

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):

$$\begin{bmatrix} 0 & 0 & 0 \\ \frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 \\ \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \\ \tilde{a}_3 \end{bmatrix} = \begin{bmatrix} \tilde{f}(s_1) \\ \tilde{f}(s_2) \\ \tilde{f}(s_3) \end{bmatrix} + \begin{bmatrix} 0.726765132 & 0.574239947 & 0.6913436 \\ 1.4142126 & 1.668453 & 2.6463 \\ 1.2805 & 1.791 & 3.0578 \end{bmatrix} \begin{bmatrix} \tilde{a}_1 \\ \tilde{a}_2 \\ \tilde{a}_3 \end{bmatrix},$$

the Extended 6×6 matrices are

$$S = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 & 0 & 0 & 0 & 0 \\ \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{\pi}{4} & (\frac{\pi}{4})^3 & (\frac{\pi}{4})^5 & 0 \\ 0 & 0 & 0 & \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \\ 0 & 0 & 0 & \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \\ 0 & 0 & 0 & \frac{\pi}{4} - 0.01 & (\frac{\pi}{4} - 0.01)^3 & (\frac{\pi}{4} - 0.01)^5 \\ 1.4142126 & 1.668453 & 2.6463 & 0 & 0 & 0 \\ 1.2805 & 1.791 & 3.0578 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1.2805 & 1.791 & 3.0578 \\ 0 & 0 & 0 & 0 & 1.2805 & 1.791 & 3.0578 \\ \end{bmatrix},$$

and the solution of Eq. (7) is:

$$\begin{bmatrix} \underline{a}_{1} \\ \underline{a}_{2} \\ \underline{a}_{3} \\ -\overline{a}_{1} \\ -\overline{a}_{2} \\ -\overline{a}_{3} \end{bmatrix} = (S-T)^{-1}F = \begin{bmatrix} 1.0112r^{2} + 1.012r \\ -0.196r^{2} - 0.196r \\ 0.0192r^{2} + 0.0192r \\ 1.012r - 3.036 \\ 0.587 - 0.196r \\ 0.00481r - 0.057 \end{bmatrix}.$$
 (29)

i.e.

$$\tilde{a}_1 = (1.0112r^2 + 1.012r, -1.012r + 3.036)$$

$$\tilde{a}_2 = (-0.196r^2 - 0.196r, -0.587 + 0.196r)$$

$$\tilde{a}_3 = (0.0192r^2 + 0.0192r, -0.00481r + 0.057).$$

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

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

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.

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Received: May 29, 2007