\_Published by European Centre for Research Training and Development UK (www.eajournals.org)

## NUMERICAL SOLUTION OF LAPLACE EQUATION IN CURVED BORDER USING FUZZY DATA BY FINITE DIFFERENCE

#### Mhassin A. A.

Faculty of Education for Pure Science, Al-Anbar University, Iraq

**ABSTRACT:** The fuzzification of Laplace equation in two dimensions are discussed with curve border. The interval of fuzzy interval can be determined. Finite difference method applied of two different grids using five points, first for initial values and the second to solve Laplace equation numerically.

**KEYWORDS:** Fuzzy membership Function (f.m.f.), Interval of Confidence, Triangular Fuzzy Number (t.f.n.),  $\alpha$  – Cuts, Five Points Finite Difference, Domain with Curve Border .

## **INTRODUCTION**

The concept of Fuzzy differential equation was first introduced by Chang Zadeh [10]. Dubois and Prade[5] has given extension principle. Raphel [8], used five points in regular domain. Here implementing five-points for finite difference method to solve Laplace equation in two variables numerically in domain with curved boundary, then fuzzified.

#### Definitions

A triangular Fuzzy number  $\mu$  is defined by three real numbers with base as the interval [a, c] and b as the vertex of triangle. The membership function are defined as follows [1,8]:

$$\mu(x)(=) \begin{cases} \frac{x-a}{b-a} ; \text{ where } a \le x \le b \\ \frac{x-c}{b-c} ; \text{ where } b \le x \le c \\ 0 ; \text{ otherwise} \end{cases}$$

The  $\alpha$ -cuts are defined by  $\Delta_L(\alpha)(=)a + \alpha(b-a)$  and  $\Delta_R(\alpha)(=)c + \alpha(b-c)$ 

#### Finite difference using to solve Laplace equation in irregular domain

The Laplace equation in two variables is defined by

$$u_{xx}(x, y) + u_{yy}(x, y) = 0 \tag{1}$$

This equation is encountered in many application, fluid mechanics, study state, electrostatics, mass transfer, and for other areas of mechanics and physics. Replacing  $u_{xx}$  and  $u_{yy}$  by the central difference formula the value of  $u(x_i, x_j)$  at any mesh point is the arithmetic mean of the values at four neighboring mesh to the left, right, above and below which is called standard five points formula Fig.1 in curved boundary, use for finding the initial data respectively as follows

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$u_{i,j} = \frac{1}{\alpha_0} \Big[ \alpha_1 u_{i-1,j} + \alpha_2 u_{i+1,j} + \alpha_3 u_{i,j-1} + \alpha_4 u_{i,j+1} \Big]$$
(2)

W

Where 
$$\alpha_1 = \frac{2}{h_1(h_1 + h_3)}$$
,  $\alpha_2 = \frac{2}{h_2(h_2 + h_4)}$ ,  $\alpha_3 = \frac{2}{h_3(h_1 + h_3)}$ ,  $\alpha_4 = \frac{2}{h_4(h_2 + h_4)}$  and  $-\alpha_0 = \sum_{i=1}^4 \alpha_i$ , where

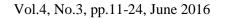
where

 $h_i \le h$  for i = 1,2,3,4 and h is the standard step length of the mesh, and if  $h_i < h$ , then  $h_i$  is the step length near the border [11].

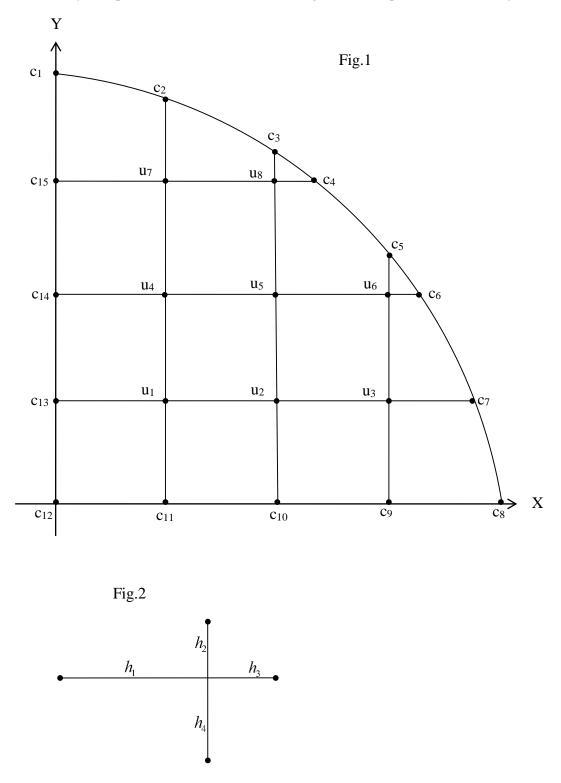
# Application of Fuzzy interval in Laplace Equation

From  $c_2$  to  $c_{15}$  represents the boundary conditions of the mesh non-square with fuzzy interval as in table 1.

Table -1-										
$c_1(=)[l_{1,1};l_{1,2};l_{1,3}]$	$c_2(=)[l_{2,1}; l_{2,2}; l_{2,3}]$	$c_3(=)[l_{3,1};l_{3,2};l_{3,3}]$	$c_4(=)[l_{4,1};l_{4,2};l_{4,3}]$							
$c_5(=)[l_{5,1};l_{5,2};l_{5,3}]$	$c_6(=)[l_{6,1}; l_{6,2}; l_{6,3}]$	$c_7(=)[l_{7,1};l_{7,2};l_{7,3}]$	$c_8(=)[l_{8,1};l_{8,2};l_{8,3}]$							
$c_9(=)[l_{9,1};l_{9,2};l_{9,3}]$	$c_{10}(=)[l_{10,1};l_{10,2};l_{10,3}]$	$c_{11}(=)[l_{11,1};l_{11,2};l_{11,3}]$	$c_{12}(=)[l_{12,1};l_{12,2};l_{12,3}]$							
$c_{13}(=)[l_{13,1};l_{13,2};l_{13,3}]$	$c_{14} = [l_{14,1}; l_{14,2}; l_{14,3}]$	$c_{15}(=)[l_{15,1};l_{15,2};l_{15,3}]$	$c_{16}(=)[l_{16,1};l_{16,2};l_{16,3}]$							



Published by European Centre for Research Training and Development UK (www.eajournals.org)



The interior points due to the non-square grid are  $u_1$  to  $u_8$ .

Now to find the initial value of  $u_5^{(0)}$  using standard five-points formula (2) as

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$u_{5}^{(0)}(=)\frac{1}{\alpha_{0}}\left[(\alpha_{1})c_{14}(+)(\alpha_{2})c_{3}(+)(\alpha_{3})c_{6}(+)(\alpha_{4})c_{10}\right]$$
(3)

We can write equation (9) in some details as

$$\left[u_{5}^{(0)}\right]^{(\alpha)}(=)\left\{\begin{array}{l} \frac{\alpha_{1}l_{14,1}+\alpha_{2}l_{3,1}+\alpha_{3}l_{6,1}+\alpha_{4}l_{10,1}}{\alpha_{0}},\\ \frac{\alpha_{1}l_{14,2}+\alpha_{2}l_{3,2}+\alpha_{3}l_{6,2}+\alpha_{4}l_{10,2}}{\alpha_{0}},\\ \frac{\alpha_{1}l_{14,3}+\alpha_{2}l_{3,3}+\alpha_{3}l_{6,3}+\alpha_{4}l_{10,3}}{\alpha_{0}},\end{array}\right\}$$

$$(4)$$

Fuzzy membership functions (f.m.f) are respective  $\alpha$  -cuts of  $c_2, c_6, c_{10}, and c_{14}$  are respectively as

Hence the  $\alpha$  – cuts of  $c_2$  is  $[c_2]^{(\alpha)} (=) [l_{2,1} + \alpha (l_{2,2} - l_{2,1}), l_{2,3} + \alpha (l_{2,2} - l_{2,3})]$ 

$$\mu_{C_{6}}(x)(=) \begin{cases} \frac{x - l_{6,1}}{l_{6,2} - l_{6,1}} ; & where \quad l_{6,1} \le x \le l_{6,2} \\ \frac{x - l_{6,3}}{l_{6,2} - l_{6,3}} ; & where \quad l_{6,2} \le x \le l_{6,3} \\ 0 & ; & otherwise \end{cases}$$
(6)

Hence the  $\alpha$  – cuts of  $c_6$  is  $[c_6]^{(\alpha)}(=)[l_{6,1} + \alpha(l_{6,2} - l_{6,1}), l_{6,3} + \alpha(l_{6,2} - l_{6,3})]$ 

$$\mu_{C_{10}}(x)(=) \begin{cases} \frac{x - l_{10,1}}{l_{10,2} - l_{10,1}} ; & where \quad l_{10,1} \le x \le l_{10,2} \\ \frac{x - l_{2,3}}{l_{10,2} - l_{10,3}} ; & where \quad l_{10,2} \le x \le l_{10,3} \\ 0 & ; & otherwise \end{cases}$$

$$(7)$$

Published by European Centre for Research Training and Development UK (www.eajournals.org) Hence the  $\alpha$  – cuts of  $c_{10}$  is  $[c_{10}]^{(\alpha)} (=) [l_{10,1} + \alpha (l_{10,2} - l_{10,1}), l_{10,3} + \alpha (l_{10,2} - l_{10,3})]$ 

$$\mu_{C_{14}}(x)(=) \begin{cases} \frac{x - l_{10,1}}{l_{10,2} - l_{10,1}} ; & where \quad l_{2,1} \le x \le l_{2,2} \\ \frac{x - l_{14,3}}{l_{14,2} - l_{14,3}} ; & where \quad l_{14,2} \le x \le l_{14,3} \\ 0 & ; & otherwise \end{cases}$$

$$(8)$$

Hence the  $\alpha$  – cuts of  $c_{14}$  is  $[c_{14}]^{(\alpha)} (=) [l_{14,1} + \alpha (l_{14,2} - l_{14,1}), l_{14,3} + \alpha (l_{14,2} - l_{14,3})]$ Then from the equation (4), the interval of the confidence for  $u_5^{(0)}$  is

$$\mu_{\mu_{5}^{(0)}}(x)(=) \begin{cases} \frac{\alpha_{1}(l_{14,2}-l_{14,1}) + \alpha_{2}(l_{2,2}-l_{2,1}) + \alpha_{3}(l_{6,2}-l_{6,1}) + \alpha_{4}(l_{10,2}-l_{10,2})}{\alpha_{0}} \alpha \\ + \frac{\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,1}}{\alpha_{0}}, \\ \frac{\alpha_{1}(l_{14,2}-l_{14,3}) + \alpha_{2}(l_{2,2}-l_{2,3}) + \alpha_{3}(l_{6,2}-l_{6,3}) + \alpha_{4}(l_{10,2}-l_{10,3})}{\alpha_{0}} \alpha \\ + \frac{\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3}}{\alpha_{0}} \alpha \end{cases}$$

or

$$\mu_{u_{5}^{(0)}}(x)(=) \begin{cases} \frac{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1}) + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,2})}{\alpha_{0}} \alpha_{0} \\ + \frac{\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,1}}{\alpha_{0}}, \\ \frac{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3})}{\alpha_{0}} \alpha_{0} \\ + \frac{\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3}}{\alpha_{0}} \alpha_{0} \end{cases}$$

To retain two roots with  $\alpha \in [0,1]$ , let

## Vol.4, No.3, pp.11-24, June 2016

\_\_Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\begin{split} x_{1} &= \frac{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1}) + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,2})}{\alpha_{0}} \alpha_{0} \\ &+ \frac{\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,1}}{\alpha_{0}}, \end{split}$$

and

$$x_{2} = \frac{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3})}{\alpha_{0}}\alpha_{0}$$
  
+  $\frac{\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3}}{\alpha_{0}}\alpha_{0}$ 

Hence

$$\alpha = \frac{\alpha_0 x_1 - (\alpha_1 l_{14,1} + \alpha_2 l_{2,1} + \alpha_3 l_{6,1} + \alpha_4 l_{10,1})}{(\alpha_1 l_{14,2} + \alpha_2 l_{2,2} + \alpha_3 l_{6,2} + \alpha_4 l_{10,2}) - (\alpha_1 l_{14,1} + \alpha_2 l_{2,1}) + \alpha_3 l_{6,1} + \alpha_4 l_{10,2})}$$

and

$$\alpha = \frac{\alpha_0 x_2 - (\alpha_1 l_{14,3} + \alpha_2 l_{2,3} + \alpha_3 l_{6,3} + \alpha_4 l_{10,3})}{(\alpha_1 l_{14,2} + \alpha_2 l_{2,2} + \alpha_3 l_{6,2} + \alpha_4 l_{10,2}) - (\alpha_1 l_{14,3} + \alpha_2 l_{2,3} + \alpha_3 l_{6,3} + \alpha_4 l_{10,3})}$$

$$\mu_{u_{5}^{0}} = \begin{cases} \frac{\alpha_{0}x_{1} - (\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,1})}{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,2})} \\ where \quad \frac{1}{\alpha_{0}}(\alpha_{1}l_{14,1} + \alpha_{2}l_{2,1} + \alpha_{3}l_{6,1} + \alpha_{4}l_{10,1}) \leq x \leq \frac{1}{\alpha_{0}}(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2})}{(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) - (\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3})} \\ where \quad \frac{1}{\alpha_{0}}(\alpha_{1}l_{14,2} + \alpha_{2}l_{2,2} + \alpha_{3}l_{6,2} + \alpha_{4}l_{10,2}) \leq x \leq \frac{1}{\alpha_{0}}(\alpha_{1}l_{14,3} + \alpha_{2}l_{2,3} + \alpha_{3}l_{6,3} + \alpha_{4}l_{10,3})}{0 \quad Otherwise} \end{cases}$$
(9)

Where  $\alpha \in [0,1]$ .

In similar way we can find out fuzzy membership functions for  $u_8^{(0)}, u_2^{(0)}, u_6^{(0)}, u_3^{(0)}, u_4^{(0)}, u_1^{(0)}$ and  $u_7^{(0)}$  by different mesh for these initial values, then we use standard for an irregular mesh as in fig.1.

Next successive approximations with their f.m.f. as required be obtain from previous approximations and specified boundary conditions.

<u>Published by European Centre for Research Training and Development UK (www.eajournals.org)</u> Numerical example

# Let us consider the Laplace equation [8,9],

$$u_{xx}(+)u_{yy}(=)0$$
 (10)

In the domain  $0 \le x \le 4, 0 \le y \le 4$  with boundary conditions  $u(0, y)(=)0, u(x, 0)(=)\frac{x^2}{2}$  and u(x, y)(=)2x(+)2y in the curved boundary as in fig.1. Leibmann's process will be applied to

## Solution

solve equation (10).

The boundary conditions given the numerical value of  $c_1 = 0, c_2 = 9.745968, c_3 = 10.9282, c_4 = 11.2915, c_5 = 11.2915, c_6 = 10.9282, c_7 = 9.745968, c_8 = 8, c_9 = 4.5, c_{10} = 2.0, c_{11} = 0.5, c_{12} = 0, c_{13} = 0, c_{14} = 0, \text{ and } c_{15} = 0.$ 

The initial values of  $u_i = 1, 2, 3, ..., 8$ , may be calculated the initial values with the help of fivepoints formulas to get the approximate solution. Hence interval of confidence for  $u_5^{(0)}$  is

$$\left[u_{5}^{(0)}\right]^{(\alpha)}(=)\left[0.001\alpha + 6.389 , -0.001\alpha + 6.391\right]$$
(11)

To find f.m.f. and respective interval of confidence these eight  $c_i$ 's as follows

$$\mu_{C_2}(x)(=) \begin{cases} \frac{x - 9.744}{9.745 - 9.746} ; \text{ where } 9.744 \le x \le 9.745 \\ \frac{x - 9.746}{9.45 - 9.746} ; \text{ where } 9.745 \le x \le 9.756 \\ 0 ; \text{ otherwise} \end{cases} , \quad \begin{bmatrix} c_2 \end{bmatrix}^{(\alpha)} (=) \begin{bmatrix} 0.001\alpha + 9.744, -0.001\alpha + 9.746 \end{bmatrix}$$

$$\mu_{c_{6}}(x)(=) \begin{cases} \frac{x - 10.927}{10.928 - 10.927} ; \text{ where } 10.927 \le x \le 10.928 \\ \frac{x - 10.929}{10.928 - 10.929} ; \text{ where } 10.928 \le x \le 10.929 \\ 0 ; \text{ otherwise} \end{cases} , \quad \begin{bmatrix} c_{6} \end{bmatrix}^{(\alpha)} (=) \begin{bmatrix} 0.001\alpha + 10.927, -0.001\alpha + 10.929 \end{bmatrix}$$

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\mu_{C_{10}}(x)(=) \begin{cases} \frac{x-1.999}{2-1.999} ; & where \ 1.999 \le x \le 2 \\ \frac{x-2.001}{2-2.001} ; & where \ 2 \le x \le 2.001 \\ 0 & ; & otherwise \end{cases} , \ \begin{bmatrix} c_{10} \end{bmatrix}^{(\alpha)} (=) \begin{bmatrix} 0.001\alpha + 1.999, -0.001\alpha + 2.001 \end{bmatrix}$$

$$\mu_{C_{14}}(x)(=) \begin{cases} \frac{x+0.001}{0+0.001} \; ; \; where \; -0.001 \le x \le 0 \\ \frac{x-0.001}{0-0.001} \; ; \; where \; \; 0 \le x \le 0.001 \\ 0 \; ; \; otherwise \end{cases} \;, \begin{bmatrix} c_{14} \end{bmatrix}^{(\alpha)}(=) \begin{bmatrix} 0.001\alpha - 0.001, -0.001\alpha + 0.001 \end{bmatrix}$$

Now the parameter  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ ,  $\alpha_4$ , and  $\alpha_0$ , for each points  $u_1$ ,  $u_2$ ,  $u_3$ ,  $u_4$ ,  $u_5$ ,  $u_6$ ,  $u_7$ , and  $u_8$ , can be find as in equation (2), in the following h and  $\alpha$ -table

h (for the initial guess)				$\alpha$ (for the initial guess)					
	$h_1$	$h_2$	$h_3$	$h_4$	$\alpha_1$	$lpha_{_2}$	$\alpha_{_3}$	$lpha_4$	$lpha_{_0}$
<i>u</i> <sub>5</sub>	2.0	1.4641 02	1.4641 02	2.0	0.2886 75	0.39433 7	0.3943 37	0.2886 75	-1.36602
<i>u</i> <sub>8</sub>	2.0	0.4641 02	0.6457 51	1.0	0.3779 65	2.94337	1.1706 2	1.3660 3	-5.85798
<i>u</i> <sub>2</sub>	2.0	1.0	1.8729 84	1.0	0.2581 99	1.0	0.2757 09	1.0	-2.53391
<i>u</i> <sub>6</sub>	2.0	0.6457 51	0.4641 02	2.0	1.3660 3	1.17062	2.9433 7	0.3779 65	-5.85798
<i>u</i> <sub>3</sub>	1.0	1.0	0.8729 84	1.0	1.0678 1	1.0	1.2237 8	1.0	-4.29099
<i>u</i> <sub>4</sub>	1.0	2.0	1.0	2.0	1.0	0.27570 9	1.0	0.2581 99	-2.53391
<i>u</i> <sub>7</sub>	1.0	0.8729 84	1.0	1.0	1.0	1.22318	1.0	1.0678 1	-4.29099
<i>u</i> <sub>1</sub>	1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	-4.0

<u>Published by European Centre for Research Training and Development UK (www.eajournals.org)</u> The following process to find the initial values of the point  $u_5^0$  depending on the above tables with the equation (3)

$$u_{5}^{(0)}(=)\frac{1}{\alpha_{0}}\left[(\alpha_{1})c_{14}(+)(\alpha_{2})c_{3}(+)(\alpha_{3})c_{6}(+)(\alpha_{4})c_{10}\right]$$

$$(=)\left\{\beta_{1}c_{14}(+)\beta_{2}c_{3}(+)\beta_{3}c_{6}(+)\beta_{4}c_{10}\right\}, \quad where \quad \beta_{i} = \frac{\alpha_{i}}{\alpha_{0}}, \quad i = 1, 2, 3, 4.$$

$$(12)$$

Hence we the interval of confidence for  $u_5^{(0)}$ ,  $[u_5^{(0)}]^{(\alpha)} = \left[ 0.001\alpha + 6.389 , -0.001\alpha + 6.391 \right]$ , we are to retain two roots with  $\alpha \in [0,1]$ . Let  $0.001\alpha + 6.389 = x_1$  and  $-0.001\alpha + 6.391 = x_2$ , then solving for  $\alpha$  we get

$$\alpha = \frac{x_1 - 6.389}{0.001} \text{ and } \alpha = \frac{x_2 - 6.391}{-0.001}, \text{ so f.m.f. for } u_5^{(0)} \text{ is}$$

$$\mu_{u_5^{(0)}}(x)(=) \begin{cases} \frac{x - 6.389}{6.390 - 6.389} ; \text{ where } 6.389 \le x \le 6.390 \\ \frac{x - 6.391}{6.390 - 6.391} ; \text{ where } 6.390 \le x \le 6.391 \\ 0 ; \text{ otherwise} \end{cases}$$

In the same way we find f.m.f. for  $u_8^{(0)}$ ,  $u_2^{(0)}$ ,  $u_6^{(0)}$ ,  $u_3^{(0)}$ ,  $u_4^{(0)}$ ,  $u_7^{(0)}$ , and  $u_1^{(0)}$  are respectively.

\_\_Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\mu_{u_{3}^{(0)}}(x)(=) \begin{cases} \frac{x - 7.18637}{7.18737 - 7.18637} ; where 5.049 \le x \le 7.18737 \\ \frac{x - 7.1837}{7.18737 - 7.18837} ; where 7.18737 \le x \le 7.18837 \\ 0 ; otherwise \end{cases} ,$$

In the following there are f.m.f of the first and forth approximations respectively using fivepoints by the method of Lebmann's iteration process applied to equation (4) with aid of the following tables

$$\mu_{u_{1}^{(1)}}(x)(=) \begin{cases} \frac{x-2.1799834420}{2.1809834420-2.1799834420} ; where \ 2.1799834420 \leq x \leq 2.1809834420 \\ \frac{x-2.1819834420}{2.1809834420-2.1819834420} ; where \ 2.1809834420 \leq x \leq 2.1819834420 \\ 0 ; otherwise \end{cases} , where \ 2.1809834420 \leq x \leq 2.1819834420 \\ \frac{x-4.5240998197}{4.5250998197-4.5240998197} ; where \ 4.5240998197 \leq x \leq 4.5250998197 \\ \frac{x-4.5260998197}{4.5250998197-4.5260998197} ; where \ 4.5250998197 \leq x \leq 4.5260998197 \\ 0 ; otherwise \end{cases} , where \ 4.5250998197 \leq x \leq 4.5260998197 \\ \frac{x-4.5260998197-4.5260998197}{4.5250998197-4.5260998197} ; where \ 4.5250998197 \leq x \leq 4.5260998197 \\ \end{cases}$$

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\mu_{u_{1}^{(1)}}(x)(=) \begin{cases} \frac{x - 7.1909920574 - 7.1909920574}{7.1919920574 - 7.192992074}; where 7.1909920574 \leq x \leq 7.1919920574 \\ \frac{x - 7.192992074}{7.1919920574 - 7.192992074}; where 7.1919920574 \leq x \leq 7.192992074 \\ 0 ; otherwise \end{cases}, where 7.1919920574 \leq x \leq 7.192992074 \\ \frac{x - 3.6838677914}{3.6848677914 - 3.6838677914}; where 3.6838677914 \leq x \leq 3.6848677914 \\ \frac{x - 3.6858677914 - 3.6858677914}{3.6848677914 - 3.6858677914}; where 3.6848677914 \leq x \leq 3.6848677914 \\ 0 ; otherwise \end{cases}, where 3.6848677914 \leq x \leq 3.6858677914 \\ \frac{x - 6.7825396356 - 6.7825396356}{6.7835396356 - 6.7845396356}; where 6.7825396356 \leq x \leq 6.7835396356 \\ \frac{x - 6.7835396356 - 6.7845396356}{6.7835396356 - 6.7845396356}; where 6.7825396356 \leq x \leq 6.7845396356 \\ 0 ; otherwise \end{cases}, where 9.6420649696 \leq x \leq 9.64306496966 \\ \frac{x - 9.6420649696 - 9.6420649696}{9.6430649696 - 9.6420649696}; where 9.6420649696 \leq x \leq 9.6430649696 \\ \frac{x - 9.6420649696 - 9.6420649696}{9.6430649696 - 9.6420649696}; where 9.6430649696 \leq x \leq 9.6430649696 \\ \frac{x - 5.8652966744 - 5.8652966744}{5.8662966744 - 5.8652966744}; where 5.8652966744 \leq x \leq 5.8662966744 \\ \frac{x - 5.8672966744 - 5.8652966744}{9.4255482308 - 9.4245482308 \leq x \leq 9.4255482308} \\ \mu_{u_{1}^{(0)}}(x)(=) \begin{cases} \frac{x - 9.4245482308}{9.4255482308 - 9.4255482308}; where 9.4255482308 \leq x \leq 9.4265482308 \\ \frac{x - 9.4265482308}{9.4255482308 - 9.4265482308}; where 9.4255482308 \leq x \leq 9.4265482308 \\ 0 ; otherwise \end{cases}$$

and the f.m.f of the fourth iterations as follows

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\mu_{u_{1}^{(4)}}(x)(=) \begin{cases} \frac{x-2.1946300042}{2.1956300042-2.1966300042}; where 2.1946300042 \le x \le 2.1956300042 \\ \frac{x-2.1966300042}{2.1956300042-2.1966300042}; where 2.1956300042 \le x \le 2.1966300042 \\ 0 : otherwise \end{cases}; where 2.1956300042 \le x \le 2.1966300042 \\ 0 : otherwise \end{cases}; where 4.5613018816 \le x \le 4.5623018816 \\ \frac{x-4.5623018816-4.5613018816}{4.5623018816-4.5633018816}; where 4.5623018816 \le x \le 4.5633018816 \\ 0 : otherwise \end{cases}; where 7.2117460739 \le x \le 7.2127460739 \\ \frac{x-7.2137460739-7.2117460739}{7.2127460739-7.2117460739}; where 7.2117460739 \le x \le 7.2127460739 \\ \frac{x-7.2137460739-7.2137460739}{7.2127460739-7.2137460739}; where 7.2127460739 \le x \le 7.2137460739 \\ 0 : otherwise \end{cases}; where 3.7352139746 \le x \le 3.7352139746 \\ \frac{x-3.7352139746}{3.7362139746-3.7352139746}; where 3.7352139746 \le x \le 3.7372139746 \\ 0 : otherwise \end{cases}; where 6.8486724996 \le x \le 6.8496724996 \\ \frac{x-6.8486724996}{6.8496724996-6.8486724996}; where 6.8486724996 \le x \le 6.8506724996 \\ \frac{x-6.8506724996}{6.8496724996-6.8486724996}; where 9.6576674056 \le x \le 9.6586674056 \\ \frac{x-9.6596674056}{9.6586674056-9.6596674056}; where 9.6586674056 \le x \le 9.6596674056 \\ 0 : otherwise \end{cases};$$

Vol.4, No.3, pp.11-24, June 2016

Published by European Centre for Research Training and Development UK (www.eajournals.org)

$$\mu_{u_{7}^{(4)}}(x)(=) \begin{cases} \frac{x-5.9075901512}{5.9085901512-5.9075901512} ; where 5.9075901512 \le x \le 5.9085901512 \\ \frac{x-5.9095901512}{5.9085901512-5.9095901512} ; where 5.9085901512 \le x \le 5.9095901512 \\ 0 ; otherwise \end{cases} \text{ and } \\ \\ \mu_{u_{8}^{(4)}}(x)(=) \begin{cases} \frac{x-9.4436848069}{9.4446848069-9.4436848069} ; where 9.4436848069 \le x \le 9.4446848069 \\ \frac{x-9.4456848069-9.4436848069}{9.4446848069-9.4456848069} ; where 9.4446848069 \le x \le 9.4456848069 \\ 0 ; otherwise \end{cases}$$

## CONCLUSION

For the given initial values the fourth approximations to solve the above example numerically is significant results in the domain has curved border. Using five points only, however may increased the accuracy as desired if we take more iterations.

## REFERENCE

- [1] Baruah, Hemanta K. Set Superimposition and its applications to the theory of Fuzzy sets. Journal of Assam Science Society, Vol. 40. Nos. 1 & 2, 25-31 (1999).
- [2] Baruah, Hemanta K. Construction of the membership function of Fuzzy number, ICIC express letters (2010a).
- [3] Baruah, Hemanta K. The mathematics of Fuzziness Myths and realities, Lambert Academic Publishing, Saarbruken, Germany (2010b).
- [4] Chang, S. L., Zadeh, L. A. On Fuzzy mapping and control, IEEE Trans. Systems man cyber net, 2 (1972) pp.30-34.
- [5] Dubois, D., Prade, H. Towards Fuzzy differential calculus, Fuzzy Sets and system, Part 3, 8 (1982) pp. 225-233.
- [6] Grewal, B. S. Numerical methods in engineering and science with program in C & C++, Khanna Publishers, New Delhi-110002,(2010) pp. 343-348.
- [7] Kaufmann, A. and Gupta, M.M. Introduction to Fuzzy arithmetic. Theory and applications, Van Nostrand Reinhold Co. Inc., Wokingham, Berkshire (1984).
- [8] Mhassn, A. A. Numerical Solution of Laplace Equation Using Fuzzy Data By Nine Points Finite Difference, IJMA -5(6), 239-246, 2014.
- [9] Raphel Kumar Saikia, Numerical Solution of Laplace Equation using Fuzzy Data, International Journal of computer applications (0975-8887) Vol. 38, No. 8, January, (2012).
- [10] Sastry, S.S. Introductory methods of numerical analysis, PHI Learning private limited, New Delhi 110001, (2009).
- [11] Young, D. M. A Survey of Numerical Mathematics, Vol. II, Addison-Wesley, 1973.
- [12] Zadeh, L. A. Probability measure of Fuzzy events, Journal Mathematical Analysis and

Published by European Centre for Research Training and Development UK (www.eajournals.org)

Applications, Vol. 23 No.2, August, (1968) pp. 421-427.