### General Linear Method for Solving Various Types of Differential Equations

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The idea of General Linear Methods came from the generalizations of traditional methods and it appears in Butcher (1985). He proposed that the generalization of both multivalue multistep method and multistage Runge-Kutta method in a natural way leads to GLM. A comprehensive study on GLM later published in Butcher (2006, 2008).



### General Linear Method General Formula

Consider the first order ordinary differential equation:

$$y'(t) = f(t, y(t)), \qquad y(t_0) = y_0$$

The general formula of General Linear Method:

$$Y_{i} = \sum_{j=1}^{s} a_{ij}hF_{j} + \sum_{j=1}^{r} u_{ij}y_{j}^{[n-1]}, \qquad i = 1, 2, \dots, s,$$
$$y_{i}^{[n]} = \sum_{j=1}^{s} b_{ij}hF_{j} + \sum_{j=1}^{r} v_{ij}y_{j}^{[n-1]}, \qquad i = 1, 2, \dots, r.$$

Coefficients of GLM could be represented in Butcher's Table:

$$\begin{array}{c|c|c} A_{s \times s} & U_{s \times r} \\ \hline B_{r \times s} & V_{r \times r} \end{array}$$

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- To obtain values of coefficients for GLM, we need to solve several order conditions.
- Derivation of order conditions is based on two equations given:

$$\eta(t) = A(\eta D)(t) + U\xi(t)$$
$$E\xi(t) = B(\eta D)(t) + V\xi(t)$$

where  $\xi$  represents the input approximation computed by a starting method, the stage values are denoted by  $\eta$  and the stage derivatives are represented by  $\eta D$ .



Matrix representation of coefficients for GLM with s = 3, r = 2:

$$\begin{array}{c|c} A = \begin{bmatrix} 0 & 0 & 0 \\ a_{21} & 0 & 0 \\ a_{31} & a_{32} & 0 \end{bmatrix} & U = \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \\ u_{31} & u_{32} \end{bmatrix} \\ \hline B = \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{bmatrix} & V = \begin{bmatrix} 1 & v_{12} \\ v_{21} & 0 \end{bmatrix}$$



### General Linear Method Values of $\xi(t)$ and $E\xi(t)$ for $t \in T^{\#}$

Before applying Theorem 2.1, we outset the values of  $\xi(t) = \begin{bmatrix} \xi_1 \\ \xi_2 \end{bmatrix}$  and

$$E\xi(t) = \begin{bmatrix} E\xi_1 \\ E\xi_2 \end{bmatrix}$$
 for  $T^{\#} = \{t_0, t_1, \dots, t_8\}$ :

t	$t_0$	$t_1$	$t_2$	$t_3$	$t_4$	$t_5$	$t_6$	$t_7$	$t_8$
tree	Ø	•	1	V	ł	¥	$\checkmark$	Y	ŧ
$\xi_1$	1	0	0	0	0	0	0	0	0
$\xi_2$	1	$^{-1}$	$\xi_{22}$	$\xi_{23}$	$\xi_{24}$	$\xi_{25}$	$\xi_{26}$	$\xi_{27}$	$\xi_{28}$
$E\xi_1$	1	1	$\frac{1}{2}$	$\frac{1}{3}$	$\frac{1}{6}$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{12}$	$\frac{1}{24}$
$E\xi_2$	1	-1	0	0	0	0	0	0	0



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### General Linear Method Operation of partitioning $(\eta D)(t) \rightarrow \eta(t)$





# General Linear Method Calculation of $\eta(t)$

$$\begin{split} \eta(\emptyset) &= A(\eta D)(\emptyset) + U\xi(\emptyset) \\ &= \begin{bmatrix} 0 & 0 & 0 \\ a_{21} & 0 & 0 \\ a_{31} & a_{32} & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \\ u_{31} & u_{32} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} u_{11} + u_{12} \\ u_{21} + u_{22} \\ u_{31} + u_{32} \end{bmatrix}. \end{split}$$

$$\begin{split} \eta(t_1) &= A(\eta D)(t_1) + U\xi(t_1) \\ &= A\eta(t_0) + U\xi(t_1) \\ &= \begin{bmatrix} 0 & 0 & 0 \\ a_{21} & 0 & 0 \\ a_{31} & a_{32} & 0 \end{bmatrix} \begin{bmatrix} u_{11} + u_{12} \\ u_{21} + u_{22} \\ u_{31} + u_{32} \end{bmatrix} + \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \\ u_{31} & u_{32} \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} -u_{12} \\ a_{21}(u_{11} + u_{12}) - u_{22} \\ a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32} \end{bmatrix}. \end{split}$$

(1)

(2)

# General Linear Method Calculation of $(E\xi)(t)$

$$(E\xi)(\emptyset) = B(\eta D)(\emptyset) + V\xi(\emptyset) \begin{bmatrix} 1\\1 \end{bmatrix} = \begin{bmatrix} b_{11} & b_{12} & b_{13}\\b_{21} & b_{22} & b_{23} \end{bmatrix} \begin{bmatrix} 0\\0\\0 \end{bmatrix} + \begin{bmatrix} 1 & v_{12}\\v_{21} & 0 \end{bmatrix} \begin{bmatrix} 1\\1 \end{bmatrix} = \begin{bmatrix} 1+v_{12}\\v_{21} \end{bmatrix}.$$

$$\begin{split} (E\xi)(t_1) &= B(\eta D)(t_1) + V\xi(t_1) \\ \begin{bmatrix} 1 \\ 0 \end{bmatrix} &= B\eta(t_0) + V\xi(t_1) \\ &= \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \\ u_{31} & u_{32} \end{bmatrix} + \begin{bmatrix} 1 & v_{12} \\ v_{21} & 0 \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} \\ &= \begin{bmatrix} b_{11}(u_{11} + u_{12}) + b_{12}(u_{21} + u_{22}) + b_{13}(u_{31} + u_{32}) - v_{12} \\ b_{21}(u_{11} + u_{12}) + b_{22}(u_{21} + u_{22}) + b_{23}(u_{31} + u_{32}) \end{bmatrix}. \end{split}$$

(3)

(4)

#### Order conditions for third order GLM wih s = 3, r = 2

No	Order conditions
1	$b_{11}(u_{11}+u_{12})+b_{12}(u_{21}+u_{22})+b_{13}(u_{31}+u_{32})-v_{12}=1$
2	$b_{21}(u_{11}+u_{12})+b_{22}(u_{21}+u_{22})+b_{23}(u_{31}+u_{32})=0$
3	$\begin{array}{l} -b_{11}u_{12}+b_{12}(a_{21}(u_{11}+u_{12})-u_{22})+b_{13}(a_{31}(u_{11}+u_{12})+a_{32}(u_{21}+u_{22})-u_{32})\\ +v_{12}\xi_{22}=\frac{1}{2}\end{array}$
4	$-b_{21}u_{12} + b_{22}(a_{21}(u_{11} + u_{12}) - u_{22}) + b_{23}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32}) = 0$
5	$ \begin{split} & b_{11}u_{12}^2 + b_{12}\big(a_{21}(u_{11}+u_{12})-u_{22}\big)^2 + b_{13}\big(a_{31}(u_{11}+u_{12})+a_{32}(u_{21}+u_{22})-u_{32}\big)^2 \\ & + v_{12}\xi_{23} = \frac{1}{3} \end{split} $
6	$b_{21}u_{12}^2 + b_{22}(a_{21}(u_{11} + u_{12}) - u_{22})^2 + b_{23}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})^2 = 0$
7	$\begin{array}{l} b_{11}u_{12}\xi_{22}+b_{12}(\xi_{22}u_{22}-a_{21}u_{12})+b_{13}\big(-a_{31}u_{12}+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22}\big)\\ +v_{12}\xi_{24}=\frac{1}{8}\end{array}$
8	$ \begin{split} b_{21} u_{12} \xi_{22} + b_{22} \left(\xi_{22} u_{22} - a_{21} u_{12}\right) + b_{23} \left(-a_{31} u_{12} + a_{32} (a_{21} (u_{11} + u_{12}) - u_{22}) \right. \\ \left. + u_{32} \xi_{22}\right) = 0 \end{split} $
9	$\begin{array}{l} -b_{11}u_{12}^3 + b_{12}(a_{21}(u_{11} + u_{12}) - u_{22})^3 + b_{13}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})^3 \\ + v_{12}\xi_{23}\frac{1}{4} = \frac{1}{4}\end{array}$
10	$\begin{array}{l} -b_{21}u_{12}^3+b_{22}(a_{21}(u_{11}+u_{12})-u_{22})^3+b_{23}(a_{31}(u_{11}+u_{12})+a_{32}(u_{21}+u_{22})\\ -u_{32})^3=0 \end{array}$
11	$\begin{array}{l} -b_{11}u_{12}^2\xi_{22}+b_{12}(a_{21}(u_{11}+u_{12})-u_{22})(\xi_{22}u_{22}-a_{21}u_{12})+b_{13}(a_{31}(u_{11}+u_{12})+a_{32}(u_{21}+u_{22})-u_{32})(-a_{31}u_{12}+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22})+v_{12}\xi_{26}=\frac{1}{2}\\ +u_{22}(a_{21}(u_{21}+u_{22})-u_{22})(-a_{31}(u_{21}+u_{22})+u_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22})+v_{12}\xi_{26}=\frac{1}{2}\end{array}$
12	$-b_{21}u_{12}^2\xi_{22} + b_{22}(a_{21}(u_{11} + u_{12}) - u_{22})(\xi_{22}u_{22} - a_{21}u_{12}) + b_{23}(a_{31}(u_{11} + u_{12}))$
	$+a_{32}(u_{21}+u_{22})-u_{32}(-a_{31}u_{12}+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22})=0$
13	$ b_{11}u_{12}\xi_{23} + b_{12}(a_{21}u_{12}^2 + \xi_{23}u_{22}) + b_{13}(a_{31}u_{12}^2 + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})^2 + u_{32}\xi_{23}) \\ + v_{12}\xi_{27} = \frac{1}{2} $
14	$b_{21}u_{12}\xi_{23} + b_{22}(a_{21}u_{12}^2 + \xi_{23}u_{22}) + b_{23}(a_{31}u_{12}^2 + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})^2 + u_{32}\xi_{33} = 0$
15	$\begin{array}{l}b_{11}u_{12}\xi_{22}+b_{12}(\xi_{22}a_{21}u_{12}+\xi_{24}u_{22})+b_{13}\left(a_{31}u_{12}\xi_{22}+a_{32}(\xi_{22}u_{22}-a_{21}u_{12})+u_{32}\xi_{24}\right)\\+v_{14}\xi_{38}=\frac{1}{4},\end{array}$
16	$\begin{array}{l} & & & & & & & & & \\ & & & & & & & \\ b_{21}u_{12}\xi_{24}+b_{22}(\xi_{22}a_{21}u_{12}+\xi_{24}u_{22})+b_{23}\left(a_{31}u_{12}\xi_{22}+a_{32}(\xi_{22}u_{22}-a_{21}u_{12})\right.\\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ \end{array}$

### General Linear Method Order conditions of GLM

#### Order conditions for fourth order GLM wih s = 3, r = 2

No	Order conditions
1	$\frac{b_{11}u_{12}^4 + b_{12}(a_{21}(u_{11} + u_{12}) - u_{22})^4 + b_{13}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})^4}{+ v_{10} \varepsilon_{20}}$
2	$b_{21}u_{12}^4 + b_{22}(a_{21}(u_{11} + u_{12}) - u_{22})^4 + b_{23}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})^4 = 0$
3	$b_{11}u_{12}^{3}\xi_{22} + b_{12}(a_{21}(u_{11}+u_{12})-u_{22})^{2}(-a_{21}u_{12}+u_{22}\xi_{22}) + b_{13}(a_{31}(u_{11}+u_{12}))$
	$+a_{32}(u_{21}+u_{22})-u_{32})^2(-a_{31}u_{12}+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22})+v_{12}\xi_{210}=\frac{5}{48}$
4	$b_{21}u_{12}{}^{3}\xi_{22} + b_{22}(a_{21}(u_{11} + u_{12}) - u_{22})^{2}(-a_{21}u_{12} + u_{22}\xi_{22}) + b_{23}(a_{31}(u_{11} + u_{12}))^{2}(-a_{21}u_{12} + u_{22}\xi_{22}) + b_{23}(a_{31}(u_{11} + u_{12}))^{2}(-a_{31}u_{12} + u_{32}\xi_{22}) + b_{33}(a_{31}(u_{11} + u_{12}))^{2}(-a_{31}u_{12} + u_{32})^{2}(-a_{31}u_{12} + u_{32})^{2}(-a_{31}u_{12$
	$+a_{32}(u_{21}+u_{22})-u_{32})^2(-a_{31}u_{12}+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})+u_{32}\xi_{22})=0$
5	$-b_{11}u_{12}{}^2\xi_{23}+b_{12}(a_{21}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}{}^2+u_{22}\xi_{23})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}{}^2+u_{22}\xi_{23})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}{}^2+u_{32}\xi_{23})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}{}^2+u_{32}\xi_{23})+b_{13}(a_{31}(u_{11}+u_{12})-u_{32})(a_{31}u_{12}{}^2+u_{32}\xi_{33})+b_{13}(a_{31}(u_{11}+u_{12})-u_{33})(a_{31}u_{12}{}^2+u_{33})(a_{31}u_{13}{}^2+u_{33})(a_{31}u_{12}{}^2+u_{33})(a_{31}u_{12}{}^2+u_{33})(a_{31}u_{12}{}^2+u_{33})(a_{31}u_{13}{}^2+u$
	$+a_{32}(u_{21}+u_{22})-u_{32})(a_{31}u_{12}^2+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})^2+u_{32}\xi_{23})+v_{12}\xi_{211}=\frac{1}{9}$
6	$-b_{21}u_{12}^{2}\xi_{23}+b_{22}(a_{21}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}^{2}+u_{22}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}^{2}+u_{22}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}^{2}+u_{22}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}^{2}+u_{22}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}^{2}+u_{22}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{12}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{12}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{12}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{12}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23}\xi_{23})+b_{23}(a_{31}(u_{11}+u_{12})-u_{23})(a_{31}u_{13}^{2}+u_{23})(a_{31}u_{13}^$
	$+a_{32}(u_{21}+u_{22})-u_{32})(a_{31}u_{12}^2+a_{32}(a_{21}(u_{11}+u_{12})-u_{22})^2+u_{32}\xi_{23})=0$
7	$-b_{11}u_{12}^2\xi_{24}+b_{12}(a_{21}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{12})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{13}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}+u_{22})+b_{13}(a_{21}u_{12}+u_{12})+b_{13}(a_{21}u_{12$
	$+a_{32}(u_{21}+u_{22})-u_{32})(a_{31}u_{12}\xi_{22}+a_{32}(-a_{21}u_{12}+u_{22}\xi_{22})+u_{32}\xi_{24})+v_{12}\xi_{212}=\frac{1}{18}$
8	$-b_{21}u_{12}{}^2\xi_{24}+b_{22}(a_{21}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{21}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{22}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}\xi_{24}+u_{22}\xi_{24})+b_{23}(a_{31}(u_{11}+u_{12})-u_{22})(a_{31}u_{12}+u_{22})(a_{31}u_{12}+u_{22})(a_{31}u_{12}+u_{12})+b_{32}(a_{31}u_{12}+u_{12})+b_{32}(a_{31}u_{12}+u_{12})(a_{31}u_{12}+u_{12})+b_{32}(a_{31}u_{12}+u_{12})+b_{33}(a_$
	$+a_{32}(u_{21}+u_{22})-u_{32})(a_{31}u_{12}\xi_{22}+a_{32}(-a_{21}u_{12}+u_{22}\xi_{22})+u_{32}\xi_{24})=0$
9	$b_{11}u_{12}{}^2\xi_{22}{}^2 + b_{12}(-a_{21}u_{12} + u_{22}\xi_{22})^2 + b_{13}(-a_{31}u_{12} + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})$
	$+u_{32}\xi_{22})^2 + v_{12}\xi_{213} = \frac{5}{96}$
10	$b_{21}u_{12}^{2}\xi_{22}^{2} + b_{22}(-a_{21}u_{12} + u_{22}\xi_{22})^{2} + b_{23}(-a_{31}u_{12} + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})$
	$+u_{32}\xi_{22})^2=0$
11	$b_{11}u_{12}\xi_{25} + b_{12}(-a_{21}u_{12}^3 + u_{22}\xi_{25}) + b_{13}(-a_{31}u_{12}^3 + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})^3$
	$+u_{32}\xi_{25}) + v_{12}\xi_{214} = \frac{1}{24}$
12	$b_{21}u_{12}\xi_{25} + b_{22}(-a_{21}u_{12}^3 + u_{22}\xi_{25}) + b_{23}(-a_{31}u_{12}^3 + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})^3$
	$+u_{32}\xi_{25})=0$
13	$b_{11}u_{12}\xi_{26} + b_{12}(-a_{21}u_{12}^2\xi_{22} + u_{22}\xi_{26}) + b_{13}(-a_{31}u_{12}^2\xi_{22} + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})$
	$(-a_{21}u_{12} + u_{22}\xi_{22}) + u_{32}\xi_{26}) + v_{12}\xi_{215} = \frac{1}{48}$
14	$b_{21}u_{12}\xi_{26} + b_{22}(-a_{21}u_{12}^2\xi_{22} + u_{22}\xi_{26}) + b_{23}(-a_{31}u_{12}^2\xi_{22} + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})$
	$(-a_{21}u_{12}+u_{22}\xi_{22})+u_{32}\xi_{26})=0$
15	$b_{11}u_{12}\xi_{27} + b_{12}(a_{21}u_{12}\xi_{23} + u_{22}\xi_{27}) + b_{13}(a_{31}u_{12}\xi_{23} + a_{32}(a_{21}u_{12}^2 + u_{22}\xi_{23}) + u_{32}\xi_{27})$
	$+\nu_{12}\xi_{216} = -\frac{1}{36}$
16	$b_{21}u_{12}\xi_{27} + b_{22}(a_{21}u_{12}\xi_{23} + u_{22}\xi_{27}) + b_{23}(a_{31}u_{12}\xi_{23} + a_{32}(a_{21}u_{12}^2 + u_{22}\xi_{23}) + u_{32}\xi_{27}) = 0$
17	$b_{11}u_{12}\xi_{28} + b_{12}(a_{21}u_{12}\xi_{24} + u_{22}\xi_{28}) + b_{13}(a_{31}u_{12}\xi_{24} + a_{32}(a_{21}u_{12}\xi_{22} + u_{22}\xi_{24}) + u_{32}\xi_{28})$

Calculation of order for GLM

• Using the order conditions given in before, for  $t_0 \dots t_3$  and minimizing the error norm of GLM of order four (for trees  $t_4 \dots t_8$ ) the GLM of order three is derived. In this way, we set the values  $b_{11} = \frac{1}{6}, b_{12} = \frac{2}{3}, b_{13} = \frac{1}{6}, u_{11} = 1, u_{12} = 0, u_{21} = \frac{7}{9}, u_{12} = \frac{2}{9}$  and substitute the values of  $\xi$  and  $E\xi$ .



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- The coefficients of three stages third order GLM method is shown in Table 5. The calculation of various quantities assigned to obtain the method can be found in Table 1.

Table 1: Coefficients set of third order GLM



Calculation of order for GLM

Calculation of order for GLM										
t	ξ1	ξ2	$\eta_1$	$\eta_1 D$	$\eta_2$	$\eta_2 D$	$\eta_3$	$\eta_3 D$	$E\xi_1$	$E\xi_2$
$t_1$	0	-1	0	1	$\frac{1}{2}$	1	1	1	1	0
$t_2$	0	$\frac{9}{16}$	0	0	$\frac{1}{8}$	$\frac{1}{2}$	$\frac{1}{2}$	1	$\frac{1}{2}$	0
$t_3$	0	$\frac{-1}{2}$	0	0	$\frac{-2}{27}$	$\frac{1}{4}$	$\frac{43}{54}$	1	$\frac{1}{2}$	0
$t_4$	0	$\frac{-1}{6}$	0	0	$\frac{-1}{27}$	$\frac{1}{8}$	43	$\frac{1}{2}$	$\frac{1}{6}$	0
t <sub>5</sub>	0	$\frac{1}{4}$	0	0	$\frac{1}{18}$	$\frac{1}{8}$	$\frac{100}{36}$	1	$\frac{1}{4}$	0
$t_6$	0	$\frac{1}{2}$	0	0	$\frac{1}{26}$	$\frac{1}{1}$	$\frac{1}{72}$	$\frac{1}{2}$	$\frac{1}{2}$	0
$t_7$	0	$\frac{8}{1}$	0	0	$\frac{36}{54}$	$\frac{16}{-2}$	$\frac{-2}{-2}$	$\frac{43}{54}$	$\frac{8}{1}$	0
t <sub>8</sub>	0	$\frac{12}{24}$	0	0	$\frac{54}{1}$	$\frac{\frac{27}{-1}}{27}$	$\frac{-1}{0}$	43 109	$\frac{12}{24}$	0
	1	2 <b>T</b>			100	41	2	100	<b>4T</b>	

By solving the order conditions, we obtain

Table 2a. Coefficients of GLM Set 1

$c_1 = 0$				$u_{11} = 1$	$u_{12} = 0$
$c_2 = \frac{1}{2}$	$a_{21} = \frac{13}{18}$			$u_{21} = \frac{7}{9}$	$u_{22} = \frac{2}{9}$
$c_3 = 1$	$a_{31} = \frac{-17}{9}$	$a_{32} = 2$		$u_{31} = \frac{17}{9}$	$u_{32} = \frac{-8}{9}$
	$b_{11} = \frac{1}{6}$	$b_{12} = \frac{2}{3}$	$b_{13} = \frac{1}{6}$	$v_{11} = 1$	$v_{12} = 0$
	$b_{21} = 0$	$b_{22} = 0$	$b_{23} = 0$	$v_{21} = 1$	$v_{22} = 0$



By solving the order conditions, we obtain

Table 2b. Coefficients of GLM Set 2

$c_1 = 0$				$u_{11} = 1$	$u_{12} = 0$
$c_2 = \frac{1}{2}$	$a_{21} = \frac{5}{6}$			$u_{21} = \frac{2}{3}$	$u_{22} = \frac{1}{3}$
$c_3 = 1$	$a_{31} = \frac{-7}{3}$	$a_{32} = 2$		$u_{31} = \frac{7}{3}$	$u_{32} = \frac{-4}{3}$
	$b_{11} = \frac{1}{6}$	$b_{12} = \frac{2}{3}$	$b_{13} = \frac{1}{6}$	$v_{11} = 1$	$v_{12} = 0$
	$b_{21} = 0$	$b_{22} = 0$	$b_{23} = 0$	$v_{21} = 1$	$v_{22} = 0$



By solving the order conditions, we obtain

Table 2c. Coefficients of GLM Set 3

$c_1 = 0$				$u_{11} = 1$	$u_{12} = 0$
$c_2 = \frac{1}{2}$	$a_{21} = \frac{2}{3}$			$u_{21} = \frac{5}{6}$	$u_{22} = \frac{1}{6}$
$c_3 = 1$	$a_{31} = \frac{-5}{3}$	$a_{32} = 2$		$u_{31} = \frac{5}{3}$	$u_{32} = \frac{-2}{3}$
	$b_{11} = \frac{1}{6}$ $b_{21} = 0$	$b_{12} = \frac{2}{3}$ $b_{22} = 0$	$b_{13} = \frac{1}{6}$ $b_{23} = 0$	$v_{11} = 1$ $v_{21} = 1$	$v_{12} = 0$ $v_{22} = 0$



The principal error norm is expressed as:

$$||\tau^{p+1}||_2 = \sqrt{\sum_{j=1}^{n_{p+1}} \left(\tau_j^{(p+1)}\right)^2}$$
(5)

The equations of order condition for fourth order GLM which are not satisfied are numbered (5), (7), (15) and (17) given as:

$$\tau_{5}^{4} :- b_{11}u_{12}^{2}\xi_{23} + b_{12}(a_{21}(u_{11} + u_{12}) - u_{22})(a_{21}u_{12}^{2} + \xi_{23}u_{22}) + b_{13}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})(a_{31}u_{12}^{2} + a_{32}(a_{21}(u_{11} + u_{12}) - u_{22})^{2} + u_{32}\xi_{23}) + v_{12}\xi_{211} =$$
(6)

$$\tau_7^4 :- b_{11}u_{12}^2\xi_{24} + b_{12}(a_{21}(u_{11} + u_{12}) - u_{22})(\xi_{22}a_{21}u_{12} + E_{24}u_{22}) + b_{13}(a_{31}(u_{11} + u_{12}) + a_{32}(u_{21} + u_{22}) - u_{32})(a_{31}u_{12}\xi_{22} + a_{32}(\xi_{22}u_{22} - a_{21}u_{12}) + u_{32}\xi_{24}) + v_{12}\xi_{212} = \frac{1}{18}$$
(7)

# General Linear Method Error Norm

$$\tau_{15}^{4}: b_{11}u_{12}\xi_{27} + b_{12}(\xi_{23}a_{21}u_{12} + \xi_{27}u_{22}) + b_{13}(a_{31}u_{12}\xi_{23} + a_{32}(a_{21}u_{12}^{2} + \xi_{23}u_{22}) + u_{32}\xi_{23} + v_{12}\xi_{216} = -\frac{1}{36}$$
(8)

$$\tau_{17}^{4}: b_{11}u_{12}\xi_{28} + b_{12}(\xi_{24}a_{21}u_{12} + \xi_{28}u_{22}) + b_{13}(a_{31}u_{12}\xi_{24} + a_{32}(\xi_{22}a_{21}u_{12} + \xi_{24}u_{22}) + u_{34}u_{12}\xi_{217} = -\frac{1}{72}$$
(9)

Substituting coefficients Set 1 and assuming  $\xi_{23} = -\frac{1}{3}$ ,  $\xi_{24} = -\frac{1}{6}$  into equations 6 - 9, the principal error norm for Set 1 using equation 5 is given as

$$||\tau^{4}||_{2} = \sqrt{\left(\tau_{5}^{4^{2}} + \tau_{7}^{4^{2}} + \tau_{15}^{4^{2}} + \tau_{17}^{4^{2}}\right)}$$
  
= 0.004880058119. (10)

Error Norm

Substituting coefficients Set 2 and assuming  $\xi_{23} = -\frac{1}{3}$ ,  $\xi_{24} = -\frac{1}{6}$  into equations 6 - 9, the principal error norm for Set 2 using equation 5 is given as

$$||\tau^{4}||_{2} = \sqrt{\left(\tau_{5}^{4^{2}} + \tau_{7}^{4^{2}} + \tau_{15}^{4^{2}} + \tau_{17}^{4^{2}}\right)}$$
  
= 0.01464017435. (11)

Substituting coefficients Set 3 and assuming  $\xi_{23} = -\frac{1}{3}$ ,  $\xi_{24} = -\frac{1}{6}$  into equations 6 - 9, the principal error norm for Set 3 using equation 5 is given as

$$||\tau^{4}||_{2} = \sqrt{\left(\tau_{5}^{4^{2}} + \tau_{7}^{4^{2}} + \tau_{15}^{4^{2}} + \tau_{17}^{4^{2}}\right)}$$
  
= 0.01464017435. (12)

By comparing the obtained values of error norm in 10 - 12, Set 1 of coefficients of GLM performed smaller error norm compared with two other sets.

h	: step size
MAXE	: maximum error
ΤS	: total steps taken
FCN	: total function calls
GLM1	: third order GLM Set 1
GLM2	: third order GLM Set 2
GLM3	: third order GLM Set 3
RK(3)	: third order calssical Ru
	Butcher (2008)



Runge-Kutta method from

Problem 1.1

$$y' = -y$$
  
$$y(0) = 1, \qquad 0 \le x \le 10$$

Solution :  $y(x) = e^{-x}$ Source : Zanariah and Suleiman (2011)



### GLM for ODEs Test Problems

• Problem 1.2

$$y_1'' = -y_1(y_1^2 + y_2^2)^{-\frac{3}{2}}, y_2'' = -y_2(y_1^2 + y_2^2)^{-\frac{3}{2}},$$
  
$$y_1(0) = 1, y_2(0) = 0, y_1'(0) = 0, y_2'(0) = 1, \ 0 \le x \le 1.$$

The first order scheme is

$$y_{1} = y_{3},$$
  

$$y_{2} = y_{4},$$
  

$$y_{3} = -y_{1}(y_{1}^{2} + y_{2}^{2})^{-\frac{3}{2}},$$
  

$$y_{4} = -y_{2}(y_{1}^{2} + y_{2}^{2})^{-\frac{3}{2}},$$
  

$$y_{1}(0) = 1, y_{2}(0) = 0, y_{3}(0) = 0, y_{4}(0) = 1, 0 \le x \le 1.$$
  
Solution :  

$$y_{1}(x) = \cos(x), y_{2}(x) = \sin(x), y_{3}(x) = -\sin(x), y_{4}(x) = \cos(x),$$
  
Source : Hull et al. (1972)

### GLM for ODEs

#### Numerical results

h	Methods	P1	P2
0.1	GLM1	1.033467(-6)	2.331140(-5)
	GLM2	4.924486(-6)	4.489148(-5)
	GLM3	6.743934(-6)	4.909702(-5)
	RK(3)	1.660682(-5)	1.063867(-4)
0.05	GLM1	1.735809(-7)	2.321687(-6)
	GLM2	6.286938(-7)	5.211417(-6)
	GLM3	7.365140(-7)	5.102407(-6)
	RK(3)	1.994295(-6)	1.363147(-5)
0.01	GLM1	1.643349(-9)	1.378402(-8)
	GLM2	5.095354(-9)	3.827213(-8)
	GLM3	5.260627(-9)	3.542277(-8)
	RK(3)	1.545145(-8)	1.112752(-7)
0.005	GLM1	2.091770(-10)	1.639213(-9)
	GLM2	6.378138(-10)	4.722891(-9)
	GLM3	6.480860(-10)	4.542803(-9)
	RK(3)	1.923719(-9)	1.394461(-8)
0.001	GLM1	1.697234(-12)	1.256535(-11)
	GLM2	5.108070(-12)	3.738154(-11)
	GLM3	5.130000(-12)	3.709081(-11)
	RK(3)	1.534058(-11)	1.117826(-10)

### Graph for Problem 1



Figure 1. Graph of fuction evaluation versus maximum error for Problem 1



### Graph for Problem 1



Figure 2. Graph of fuction evaluation versus maximum error for Problem 2

- In this section, we derived the order conditions and gave the 3 new sets of coefficients for the third order General Linear Method.
- Then we applied the GLM on some test problems and compared the results with RK method.
- Numerical results showed that the GLM is more accurate than RK method in solving ODEs.



### General Linear Method for Solving Volterra Integro-Differential Equtaions

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 Volterra integro-differential equations (VIDEs) are widely used in our industry such as in electrodynamics, oscillation theory, and biomathematics problems.

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- Integro-differential equations contain differential and integral operators in the equation.
- An integro-differential equation is an equation in which the unknown function y(x) appears under the integral sign and contains an ordinary derivative  $y^{(n)}(x)$ .
- A standard integro-differential equation is in the form of:

$$y^{(n)} = f(x) + \lambda \sum_{g(x)}^{h(x)} K(x, t) y(t) dt,$$
(1)

where g(x) and h(x) are the limit of the integration,  $\lambda$  is a constant parameter.K(x,t) is the kernel of the integro equation The function of f(x) and K(x,t) are given in advance.
### Volterra Integro-Differential Equations

• Consider the numerical solution of the second kind of volterra integro-differential (VIDEs) equation

$$y' = f(x) + \int_0^x k(x,t)y(t)dt, \ y(x_0) = y_0.$$
 (2)

 $y(x_n)$  denotes the exact value of y at  $x_n = x_0 + h$ .  $y_n$  to denote an approximation value of y at  $x_n$ .

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 In order to solve the integral part in (2), familiar numerical integration methods are used to approximate like Simpson Method I, Simpson Method II (Linz(1985)) and Lagrange interpolation polynomial (Filiz(2014)).

#### Definition 1 (Simpson's Method II)

If three-eighths rule is used at the upper end, on the points  $t_{(n-3)}, t_{(n-2)}, t_{(n-1)}, t_n$ , we get the weights n is even: as in Simpson I n is odd:  $w_{n0} = \frac{1}{3}, \quad n \ge 5$  $w_{n,2i} = \frac{2}{3}, \quad i = 1, 2, \dots, \frac{n-5}{2},$  $w_{n,n-3} = \frac{17}{24} - \frac{1}{3}\delta_{n3},$  $w_{n,n-1} = w_{n,n-2} = \frac{9}{8},$  $w_{nn} = \frac{3}{8}.$  $\delta_{ij}$  denotes the Kronecker delta  $\delta_{ij} = 0, i \ne j, \delta_{ii} = 0.$ 

#### Definition 2 (Simpson's Method I)

If three-eighths rule is applied on the points  $t_0, t_1, t_2, t_3$ , one gets the n is even:  $w_{n0} = w_{nn} = \frac{1}{3}$ , weights (for  $n \ge 2$ )  $w_{n,2i} = \frac{2}{3}, \qquad i = 0, 1, \dots, \frac{n}{2} - 1,$  $w_{n,2i+1} = \frac{4}{3}, \qquad i = 0, 1, \dots, \frac{n}{2} - 1.$ n is odd:  $w_{n0} = \frac{3}{8}$ ,  $w_{n1} = w_{n2} = \frac{9}{8}$  $w_{n3} = \frac{17}{24} - \frac{1}{3}\delta_{n3},$   $w_{n,2i} = \frac{4}{3}, \qquad i = 2, 3, \dots, \frac{n-1}{2},$   $w_{n,2i+1} = \frac{2}{3}, \qquad i = 2, 3, \dots, \frac{n-3}{2},$  $w_{nn} = \frac{1}{2}, \quad n > 5.$  $\delta_{ii}$  denotes the Kronecker delta  $\delta_{ii} = 0, i \neq j, \delta_{ii} = 0.$ 

#### Definition 3 (Lagrange Interpolation Polynomial)

Polynomial P(x) of degree  $\neq (n-1)$  that pass through the n points $(x_1, y_1 = f(x_1)), (x_2, y_2 = f(x_2)), \dots, (x_n, y_n = f(x_n))$ , and is given by

$$P(x) = \sum_{j=1}^{n} P_j(x),$$

where

$$P_j(x) = y_j \prod_{k=1, k \neq 1}^n \frac{x - x_k}{x_j - x_k}.$$

• Raftari in (2010) solve the VIDE using Homotopy perturbation method along with trapezoidal rule for the integral operator and compared it with the finite difference method.

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- Zarebnia in (2010), the author considered the numerical solutions of VIDE by means of the Sinc collocation method where Sinc methods are direct solvers to integral equations.
- Filiz in (2013) applied several Runge-Kutta methods with different order associated with trapezoidal rule and Simpson's rule to VIDE.

 Filiz (2014) used a higher order method which is the Runge-Kutta-Fehlberg method to solve the VIDE. He developed a new numerical routine for the integral part by using Lagrange interpolation and combination of various numerical quadrature rules to attain higher accuracy.

From (4), we let the integral part be

$$\int_0^x k(x,t)y(t)dt = z(t),$$
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 (4)

- The third order GLM is combined with appropriate numerical quadrature rule in order to approximate the integral part given in equation (7).
- To evaluate the integral z(x), we use the combination of composite Simpson's II rule for the interval  $[x_0, x_n]$

• Lagrange interpolation for interval  $[x_n, x_{(n+c_i)}]$  at points  $x = -1, x = 0, x = c_i$ .

- Lagrange interpolation for interval  $[x_n, x_{(n+c_i)}]$  at points  $x = -1, x = 0, x = c_i$ .
- Lagrange Interpolation:

$$c_2 = \frac{1}{2},$$
  $P(x) = -\frac{1}{72}y_1 + \frac{7}{24}y_2 + \frac{2}{9}y_3,$ 

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• The third order General Linear Method applied to approximate equation (8) on a set of interval [0, X] of equally spaced grid points  $x_0 < x_{1...} < x_N = X$  where  $0 \le n \le N$  with step size  $h = \frac{X - x_0}{V}$  may be written as:

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- The third order General Linear Method applied to approximate equation (8) on a set of interval [0, X] of equally spaced grid points  $x_0 < x_{1...} < x_N = X$  where  $0 \le n \le N$  with step size  $h = \frac{X x_0}{V}$  may be written as:
- For interval  $[x_0, x_n]$ , if n = 1, trapezoidal rule is used then

$$z_n = \frac{h}{2} \left( K(x_{(n+c_i)}, x_{(n-1)}) y_1(x_{(n-1)}) + (x_{(n+c_i)}, x_n) y_1(x_n) \right)$$

else if n is even, composite  $\frac{1}{3}$  Simpson's rule is used then

$$z_{n} = \frac{h}{3} (K(x_{n+c_{i}}, x_{0})y_{1}(x_{0}) + 2\sum_{m=1}^{\frac{n}{2}-1} K(x_{n+c_{i}}, x_{2m})y_{1}(x_{2m}) + 4\sum_{m=1}^{\frac{n}{2}} K(x_{n+c_{i}}, x_{2m-1})y_{1}(x_{2m-1}) + K(x_{n+c_{i}}, x_{n})y_{1}(x_{n})),$$

else if n is even, composite  $\frac{1}{3}$  Simpson's rule is used then

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else if n is odd, composite Simpson's II rule is used then

$$z_{n} = \frac{h}{3} (K(x_{n+c_{i}}, x_{0})y_{1}(x_{0}) + 2\sum_{m=1}^{\frac{n-3}{2}-1} K(x_{n+c_{i}}, x_{2m})y_{1}(x_{2m}) + 4\sum_{m=1}^{\frac{n-3}{2}} K(x_{n+c_{i}}, x_{2m-1})y_{1}(x_{2m-1}) + K(x_{n+c_{i}}, x_{n-3})y_{1}(x_{n-3}) + \frac{3h}{8} (K(x_{n+c_{i}}, x_{n-3})y_{1}(x_{n-3})) + 3K(x_{n+c_{i}}, x_{n-2})y_{1}(x_{n-2}) + \frac{34}{8} (K(x_{n+c_{i}}, x_{n-3})y_{1}(x_{n-3})) + \frac{3}{8} (K(x_{n+c_{i}}, x_{n-3})$$

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Therefore, General Linear Method after employing composite Simpson's rule and Lagrange interpolation is given as follows

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$$\begin{split} Y_1 &= u_{11}y_1(x_n) + u_{12}y_2(x_n), \\ Y_2 &= a_{21}hF(x_n,Y_1,z_n) + u_{21}y_1(x_n) + u_{22}y_2(x_n), \\ z_{n+\frac{1}{2}} &= z_n + h(-\frac{1}{72}K(x_{n+\frac{1}{2}},x_{n-1})y_1(x_{n-1}) - \frac{7}{24}K(x_{n+\frac{1}{2}},x_n)y_1(x_n) \\ &\quad -\frac{2}{9}K(x_{n+\frac{1}{2}},x_{n+\frac{1}{2}})y_1(x_{n+\frac{1}{2}})), \\ Y_3 &= a_{31}hF(x_n,Y_1,z_n) + a_{32}hF(x_{n+\frac{1}{2}},Y_1,z_{n\frac{1}{2}}) + u_{31}y_1(x_n) + u_{32}y_2(x_n), \\ z_{n+1} &= z_n + h(-\frac{1}{12}K(x_{n+1},x_{n-1})y_1(x_{n-1}) + \frac{2}{3}K(x_{n+1},x_n)y_1(x_n) \\ &\quad +\frac{5}{12}K(x_{n+1},x_{n+1})y_1(x_{n+1})), \end{split}$$

Therefore, General Linear Method after employing composite Simpson's rule and Lagrange interpolation is given as follows

$$\begin{split} Y_{1} &= u_{11}y_{1}(x_{n}) + u_{12}y_{2}(x_{n}), \\ Y_{2} &= a_{21}hF(x_{n},Y_{1},z_{n}) + u_{21}y_{1}(x_{n}) + u_{22}y_{2}(x_{n}), \\ z_{n+\frac{1}{2}} &= z_{n} + h(-\frac{1}{72}K(x_{n+\frac{1}{2}},x_{n-1})y_{1}(x_{n-1}) - \frac{7}{24}K(x_{n+\frac{1}{2}},x_{n})y_{1}(x_{n}) \\ &-\frac{2}{9}K(x_{n+\frac{1}{2}},x_{n+\frac{1}{2}})y_{1}(x_{n+\frac{1}{2}})), \\ Y_{3} &= a_{31}hF(x_{n},Y_{1},z_{n}) + a_{32}hF(x_{n+\frac{1}{2}},Y_{1},z_{n\frac{1}{2}}) + u_{31}y_{1}(x_{n}) + u_{32}y_{2}(x_{n}), \\ z_{n+1} &= z_{n} + h(-\frac{1}{12}K(x_{n+1},x_{n-1})y_{1}(x_{n-1}) + \frac{2}{3}K(x_{n+1},x_{n})y_{1}(x_{n}) \\ &+ \frac{5}{12}K(x_{n+1},x_{n+1})y_{1}(x_{n+1})), \\ Y_{1}(x_{n+1}) &= b_{11}hF(x_{n},Y_{1},z_{n}) + b_{12}hF(x_{n+\frac{1}{2}},Y_{2},z_{n\frac{1}{2}}) + \\ &b_{13}hF(x_{n+1},y_{3},z_{n+1}) + v_{11}y_{1}(x_{10}) + v_{12}y_{2}(x_{10}), \\ \end{array}$$

In order to illustrate the efficiency and accuracy of the method, we solved some numerical tests. The exact solution y(x) is used to estimate the global error. The following problems are solved by using the proposed method.

Below is the notation that we used in the table

- h : Stepsize
- MAXE : Maximum global error
- RK3, : Third-order classical Runge-Kutta method with 3 stages.

Problem 1:

$$y(x)' = 1 + \int_0^x y(t)dt, t \ge 0, y(0) = y_0.$$
  
 $y(x) = sinh(x).$ 

h	GLM3	RK3	
	MAXE		
0.1	$1.2347 \times 10^{-6}$	$4.7137 \times 10^{-6}$	
0.025	$9.9859 \times 10^{-9}$	$7.1772 \times 10^{-8}$	
0.01	$5.6041 \times 10^{-10}$	$4.6094 \times 10^{-9}$	
0.005	$6.7079 \times 10^{-11}$	$5.7715 \times 10^{-10}$	
0.001	$5.1485 \times 10^{-13}$	$4.6243 \times 10^{-12}$	

Table 1: Maximum global error for Problem 1

### Numerical Examples



Figure 1: Log<sub>10</sub> (Maximum Global Error) vs Function Evaluation for Problem 1

#### Problem 2:

$$y(x)' = 1 + 2x - y(x) - \int_0^x x(1+2x)e^{t(x-t)}y(t)dt, t \ge 0, y(0) = 1.$$
$$y(x) = e^{x^2}.$$

h	GLM3	RK3
	MAXE	
0.1	$3.9332 \times 10^{-6}$	$4.8141 \times 10^{-5}$
0.025	$1.4323 \times 10^{-7}$	$1.4400 \times 10^{-6}$
0.01	$1.0939 \times 10^{-8}$	$4.6094 \times 10^{-7}$
0.005	$1.4325 \times 10^{-9}$	$5.7715 \times 10^{-8}$
0.001	$1.1851 \times 10^{-11}$	$4.6243 \times 10^{-10}$

Table 2: Maximum global error for Problem 2

### Numerical Examples



Figure 2: Log<sub>10</sub>(Maximum Global Error) vs Function Evaluation for Problem 1

#### Problem 3:

$$y(x)' = \ln(1+x)(\frac{x}{2}\ln(1+x)+1) + \frac{1}{1+x} + y(x) - \int_0^x \frac{x}{t+1}y(t)dt,$$
$$y(0) = 0,$$
$$y(x) = \ln(x+1).$$

h	GLM3	RK3
	MAXE	
0.1	6.3429 ×10 <sup>-6</sup>	$3.3432 \times 10^{-5}$
0.025	$4.4142 \times 10^{-8}$	$6.3237 \times 10^{-7}$
0.01	$4.0164 \times 10^{-9}$	$4.1259 \times 10^{-8}$
0.005	$5.4245 \times 10^{-10}$	$5.1811 \times 10^{-9}$
0.001	$4.5689 \times 10^{-12}$	$4.1572 \times 10^{-11}$

Table 3: Maximum global error for Problem 3

### Numerical Examples



Figure 3: Log<sub>10</sub>(Maximum Global Error) vs Function Evaluation for Problem 1

- In this research, we derived the order conditions and gave a new set of coefficient for the third order General Linear Method.
- Handling the integral operator in Volterra integro-differential equations using Lagrange interpolation is demonstrated as well.
- Then we applied the GLM on some test problems and compared the results with RK method.
- Numerical results showed that the GLM is more accurate than RK method in solving VIDE.

# General Linear Method for solving Fuzzy Differential Equations

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• Fuzzy numbers generalize classical real numbers and a fuzzy number is a subset of the real line that has some additional properties.


- Fuzzy numbers generalize classical real numbers and a fuzzy number is a subset of the real line that has some additional properties.
- Definition 1 Let X be a nonempty set. A fuzzy set u in X is characterized by its membership function u : X → [0, 1].

Then u is a fuzzy number if it satisfies the following properties:

- 1. *u* is normal, there exists  $x_0 \in \mathbb{R}$  such that  $u(x_0) = 1$ ,
- 2. *u* is a convex fuzzy

$$\operatorname{set}(u(ts) + (1-t)r) \geq \min\{u(s), u(r)\}, \forall t \in [0,1], x, r \in \mathbb{R},$$

- 3. *u* is upper semicontinuous on  $\mathbb{R}$
- 4.  $cl\{x \in \mathbb{R} | u(x) > 0\}$  is compact, where cl denotes the closure of a subset.

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 $\mathbb{R}_F$  is called the space of fuzzy numbers. Then for  $0 \le r \le 1$  if  $u \in \mathbb{R}_F$ , the *r*-level set  $[u]_{\alpha} = \{x \in \mathbb{R} | u(x) \ge r\}$  is nonempty closed bounded interval. The *r*-level set of *u* is denoted by  $[u]_r = [u_r^-, u_r^+]$ .



- **Proposition 1** A fuzzy number u is completely determined by any pair  $u = (u^-, u^+)$  of functions  $u^-, u^+ : [0, 1] \to \mathbb{R}$ , defining the end-points of the *r*-cuts, satisfying the conditions:
  - u<sup>-</sup>: r → u<sup>-</sup><sub>r</sub> ∈ ℝ is a bounded monotonic nondecreasing left-continuous function ∀r ∈ (0, 1] and right-continuous for r = 0;
     u<sup>+</sup>: r → u<sup>+</sup><sub>r</sub> ∈ ℝ is a bounded monotonic nonincreasing left-continuous function ∀r ∈ (0, 1] and right-continuous for r = 0;
     u<sup>-</sup><sub>1</sub> ≤ u<sup>+</sup><sub>1</sub> for r = 1, which implies u<sup>-</sup><sub>r</sub> ≤ u<sup>+</sup><sub>r</sub> ∀r ∈ [0, 1].



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## **Fuzzy Differential Equations**

### Difference between Crisp Set and Fuzzy Set



The first order fuzzy initial value problem of fuzzy differential equation (FDE) defined by

$$y'(t) = f(t, y), \quad y(t_0) = y_0 \in \mathbb{R}_F,$$
 (1)

where function  $f : \mathbb{R} \times \mathbb{R}_F \to \mathbb{R}_F$  is continuous and  $y_0$  is a fuzzy number.

## B. Bede, S. and G. Gal (2005)

They introduced a new generalized concept of differentiability which extends the previous Hukuhara derivative in order to overcome the solutions of becoming fuzzier.



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## B. Bede (2013)

A comprehensive study on fuzzy logic and fuzzy sets are found in here.



Generalized Hukuhara-differentiability

**Definition 2** Let  $f : [a, b] \to \mathbb{R}_F$  and  $t_0 \in [a, b]$ . f is said to be strongly generalized Hukuhara differentiable at  $t_0$  if there exists element  $f'(t_0) \in \mathbb{R}_F$ , such that

i for all h > 0 sufficiently small,  $\exists f(t_0 + h) \ominus_H f(t_0), f(t_0) \ominus_H f(t_0 - h)$  and the limits (in the metric D)

$$\lim_{h\searrow 0}\frac{f(t_0+h)\ominus_H f(t_0)}{h} = \lim_{h\searrow 0}\frac{f(t_0)\ominus_H f(t_0-h)}{h} = f'_G(t_0),$$
(2)

is the (i)-differentiability on (a, b),

ii for all h > 0 sufficiently small,  $\exists f(t_0) \ominus_H f(t_0 + h), f(t_0 - h) \ominus_H f(t_0)$  and the limits

$$\lim_{h \searrow 0} \frac{f(t_0) \ominus_H f(t_0 + h)}{-h} = \lim_{h \searrow 0} \frac{f(t_0 - h) \ominus_H f(t_0)}{-h} = f'_G(t_0).$$
(3)

is the (ii)-differentiability on (a, b),

Generalized Hukuhara-differentiability

### Theorem 1:

Let  $f : [a, b] \to \mathbb{R}_F$  and  $x_0 \in [a, b]$  with  $f_r^-(t)$  and  $f_r^+(t)$  both differentiable at t. We say that

- if f is (i)-differentiable then  $[f'(t)]_r = [(f_r^-)'(t), (f_r^+)'(t)],$
- if f is (ii)- differentiable then  $[f'(t)]_r = [(f_r^+)'(t), (f_r^-)'(t)],$

$$\forall r \in [0,1]$$

$$\forall r \in [0,1]$$



Characterization Theorem

### Theorem 2 (Characterization theorem):

Consider the FIVP (1.2) where  $f:(a,b) imes \mathbb{R}_F o \mathbb{R}_F$  such that

- $i [f(t,y)]_r = [f_r^-(t,y_r^-,y_r^+), f_r^+(t,y_r^-,y_r^+)],$
- ii  $f_r^-$  and  $f_r^+$  are equicontinuous,
- iii there exists L > 0 such that

$$\begin{split} |f_r^-(t,y^-,y^+) - f_r^-(t,z^-,z^+)| &\leq L \max\{|y^--z^-|,|y^+-z^+|\},\\ |f_r^+(t,y^-,y^+) - f_r^+(t,z^-,z^+)| &\leq L \max\{|y^--z^-|,|y^+-z^+|\},\\ \end{split}$$
 where  $\forall r \in [0,1].$ 



Hence from theorem 1, the FIVP (1) is equivalent to the following two system of ODEs:

$$\begin{cases} (y_r^-)'(t) = f_r^-(t, y_r^-, y_r^+) = F(t, y_r^-, y_r^+), & y_r^-(0) = (y_0)_r^-, \\ (y_r^+)'(t) = f_r^+(t, y_r^-, y_r^+) = G(t, y_r^-, y_r^+), & y_r^+(0) = (y_0)_r^+, \end{cases}$$
(4)

$$\begin{cases} (y_r^-)'(t) = f_r^+(t, y_r^-, y_r^+) = G(t, y_r^-, y_r^+), & y_r^-(0) = (y_0)_r^-, \\ (y_r^+)'(t) = f_r^-(t, y_r^-, y_r^+) = F(t, y_r^-, y_r^+), & y_r^+(0) = (y_0)_r^+, \end{cases}$$
(5)

Then, for type (i)-differentiability the FIVP (1) and system of ODEs (5) are equivalent and for type (ii)-differentiability the FIVP (1) and system of ODEs (6) are equivalent.

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The fuzzy General Linear Method is represented by

$$y_i^{-}(t_{n+1};r) = \sum_{j=1}^{s=3} b_{ij} h F_j(t_n, y(t_n;r)) + \sum_{j=1}^{r=2} v_{ij} y_j^{-}(t_n;r), \quad i = 1, \ldots, r,$$
(6)

$$y_i^+(t_{n+1};r) = \sum_{j=1}^{s=3} b_{ij} h G_j(t_n, y(t_n;r)) + \sum_{j=1}^{r=2} v_{ij} y_j^+(t_n;r), \quad i = 1, \dots, r,$$
(7)



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#### where

$$Y_{1}^{-}(y(t_{n};r)) = u_{11}y_{1}^{-}(t_{n};r) + u_{12}y_{2}^{-}(t_{n};r)$$

$$Y_{1}^{-}(y(t_{n};r)) = u_{11}y_{1}^{-}(t_{n};r) + u_{12}y_{2}^{-}(t_{n};r)$$

$$Y_{2}^{-}(y(t_{n};r)) = a_{21}hF_{1}(t_{n},y(t_{n};r)) + u_{21}y_{1}^{-}(t_{n};r) + u_{22}y_{2}^{-}(t_{n};r)$$

$$Y_{2}^{+}(y(t_{n};r)) = a_{21}hG_{1}(t_{n},y(t_{n};r)) + u_{21}y_{1}^{+}(t_{n};r) + u_{22}y_{2}^{+}(t_{n};r)$$

$$Y_{3}^{-}(y(t_{n};r)) = a_{31}hF_{1}(t_{n},y(t_{n};r)) + a_{32}hF_{2}(t_{n},y(t_{n};r))$$

$$+ u_{31}y_{1}^{-}(t_{n};r) + u_{32}y_{2}^{-}(t_{n};r)$$

$$Y_{3}^{+}(y(t_{n};r)) = a_{31}hG_{1}(t_{n},y(t_{n};r)) + a_{32}hG_{2}(t_{n},y(t_{n};r))$$

$$+ u_{31}y_{1}^{+}(t_{n};r) + u_{32}y_{2}^{+}(t_{n};r)$$
(10)



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and

$$\begin{split} F_1(t_n, y(t_n; r)) &= \min \left\{ f(t_n + c_1 h, u) | u \in [Y_1^-(y(t_n; r)), Y_1^+(y(t_n; r))] \right\}, \\ G_1(t_n, y(t_n; r)) &= \max \left\{ f(t_n + c_1 h, u) | u \in [Y_1^-(y(t_n; r)), Y_1^+(y(t_n; r))] \right\}, \\ F_2(t_n, y(t_n; r)) &= \min \left\{ f(t_n + c_2 h, u) | u \in [Y_2^-(y(t_n; r)), Y_2^+(y(t_n; r))] \right\}, \\ G_2(t_n, y(t_n; r)) &= \max \left\{ f(t_n + c_2 h, u) | u \in [Y_2^-(y(t_n; r)), Y_2^+(y(t_n; r))] \right\}, \\ F_3(t_n, y(t_n; r)) &= \min \left\{ f(t_n + c_3 h, u) | u \in [Y_3^-(y(t_n; r)), Y_3^+(y(t_n; r))] \right\}, \\ G_3(t_n, y(t_n; r)) &= \max \left\{ f(t_n + c_3 h, u) | u \in [Y_3^-(y(t_n; r)), Y_3^+(y(t_n; r))] \right\}, \end{split}$$

(7) and (8) are the fuzzy GLM derived based on type (i)-differentiability.

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For under type (ii)-differentiability:

$$y_{i}^{-}(t_{n+1};r) = \sum_{j=1}^{s=3} b_{ij}hG_{j}(t_{n}, y(t_{n};r)) + \sum_{j=1}^{r=2} v_{ij}y_{j}^{-}(t_{n};r), \quad i = 1, \dots, r,$$

$$y_{i}^{+}(t_{n+1};r) = \sum_{j=1}^{s=3} b_{ij}hF_{j}(t_{n}, y(t_{n};r)) + \sum_{j=1}^{r=2} v_{ij}y_{j}^{+}(t_{n};r), \quad i = 1, \dots, r,$$
(11)

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#### where

$$Y_{1}^{-}(y(t_{n};r)) = u_{11}y_{1}^{-}(t_{n};r) + u_{12}y_{2}^{-}(t_{n};r)$$

$$Y_{1}^{+}(y(t_{n};r)) = u_{11}y_{1}^{+}(t_{n};r) + u_{12}y_{2}^{+}(t_{n};r)$$

$$Y_{2}^{-}(y(t_{n};r)) = a_{21}hG_{1}(t_{n},y(t_{n};r)) + u_{21}y_{1}^{-}(t_{n};r) + u_{22}y_{2}^{-}(t_{n};r)$$

$$Y_{2}^{+}(y(t_{n};r)) = a_{21}hF_{1}(t_{n},y(t_{n};r)) + u_{21}y_{1}^{+}(t_{n};r) + u_{22}y_{2}^{+}(t_{n};r)$$

$$Y_{3}^{-}(y(t_{n};r)) = a_{31}hG_{1}(t_{n},y(t_{n};r)) + a_{32}hG_{2}(t_{n},y(t_{n};r))$$

$$+ u_{31}y_{1}^{-}(t_{n};r) + u_{32}y_{2}^{-}(t_{n};r)$$

$$Y_{3}^{+}(y(t_{n};r)) = a_{31}hF_{1}(t_{n},y(t_{n};r)) + a_{32}hF_{2}(t_{n},y(t_{n};r))$$

$$+ u_{31}y_{1}^{+}(t_{n};r) + u_{32}y_{2}^{+}(t_{n};r).$$
(15)



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## Numerical results

#### Notations

- r  $y^{-}$   $y^{+}$   $\tilde{\gamma}^{-}$   $\tilde{\gamma}^{+}$   $E^{-}(t; r)$   $E^{+}(t; r)$  GLM(3) RK(3)
- : r-level set of fuzzy numbers
  - : left bounded of approximate solution
  - : right bounded of approximate solution
  - : left bounded of exact solution
  - : right bounded of exact solution
  - : left bounded error of approxiate solution
- $E^+(t; r)$  : right bounded error of approxiate solution
  - : fuzzy third order GLM Set 1
  - : fuzzy third order classical Runge-Kutta method from (Butcher, 2008)
- ABM(3) : Third order Adams-Bashforth-Moulton predictor-corrector from (Allahvinraloo et al., 2007)
  - ANN : Artificial neural network method from (Effati and Pakdaman, 2010)

$$egin{aligned} y'(t) &= y(t) \ y(0) &= [0.75 + 0.25r, 1.125 - 0.125r], \ 0 &\leq t \leq 1 \end{aligned}$$

The equivalent system of ODEs based on (i)-differentiability:

$$(y^{-})'(t;r) = y^{-}(t;r), \quad y^{-}(0;r) = 0.75 + 0.25r$$
  
 $(y^{+})'(t;r) = y^{+}(t;r), \quad y^{+}(0;r) = 1.125 - 0.125r$ 

Solutions :

$$ilde{Y}^{-}(t;r) = (0.75 + 0.25r)e^t$$
  
 $ilde{Y}^{+}(t;r) = (1.125 - 0.125r)e^t$ 

Source : Ma et al. (1999)



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## Numerical results

#### Numerical results for Problem 1

		<i>t</i> = 0.5		t = 1			
	GLM(3)	RK(3)	ABM(3)	GLM(3)	RK(3)	ABM(3)	ANN
r	$E^{-}(0.5; r)$	$E^{-}(0.5; r)$	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (1; r)			
0.0	3.735899(-6)	2.378365(-5)	6.323706(-5)	1.231826(-5)	7.842448(-5)	2.249512(-4)	8.895270(-5)
0.1	3.860429(-6)	2.457644(-5)	6.534496(-5)	1.272887(-5)	8.103863(-5)	2.324496(-4)	2.903725(-5)
0.2	3.984959(-6)	2.536923(-5)	6.745287(-5)	1.313947(-5)	8.365278(-5)	2.399480(-4)	4.693243(-5)
0.3	4.109488(-6)	2.616202(-5)	6.956077(-5)	1.355008(-5)	8.626693(-5)	2.474464(-4)	2.484654(-5)
0.4	4.234018(-6)	2.695481(-5)	7.166867(-5)	1.396069(-5)	8.888108(-5)	2.549447(-4)	4.739291(-6)
0.5	4.358548(-6)	2.774760(-5)	7.377657(-5)	1.437130(-5)	9.149523(-5)	2.624431(-4)	5.478406(-5)
0.6	4.483078(-6)	2.854039(-5)	7.588448(-5)	1.478191(-5)	9.410938(-5)	2.699415(-4)	2.827934(-5)
0.7	4.607608(-6)	2.933317(-5)	7.799238(-5)	1.519252(-5)	9.672353(-5)	2.774398(-4)	4.693161(-5)
0.8	4.732138(-6)	3.012596(-5)	8.010028(-5)	1.560313(-5)	9.933768(-5)	2.849382(-4)	9.712388(-5)
0.9	4.856668(-6)	3.091875(-5)	8.220818(-5)	1.601373(-5)	1.019518(-4)	2.924366(-4)	1.417971(-5)
1.0	4.981198(-6)	3.171154(-5)	8.431608(-5)	1.642434(-5)	1.045660(-4)	2.999350(-4)	5.574679(-5)
r	$E^+(0.5; r)$	E <sup>+</sup> (0.5; r)	$E^+(0.5; r)$	$E^+(1; r)$	$E^{+}(1; r)$	$E^{+}(1; r)$	E <sup>+</sup> (1; r)
0.0	5.603848(-6)	3.567548(-5)	9.485559(-5)	1.847739(-5)	1.176367(-4)	3.374268(-4)	6.003329(-5)
0.1	5.541583(-6)	3.527909(-5)	9.380164(-5)	1.827208(-5)	1.163296(-4)	3.336777(-4)	1.107843(-6)
0.2	5.479318(-6)	3.488269(-5)	9.274769(-5)	1.806678(-5)	1.150226(-4)	3.299285(-4)	2.582699(-5)
0.3	5.417053(-6)	3.448630(-5)	9.169374(-5)	1.786147(-5)	1.137155(-4)	3.261793(-4)	8.472236(-6)
0.4	5.354788(-6)	3.408991(-5)	9.063979(-5)	1.765617(-5)	1.124084(-4)	3.224301(-4)	3.384699(-5)
0.5	5.292523(-6)	3.369351(-5)	8.958584(-5)	1.745086(-5)	1.111013(-4)	3.186809(-4)	1.443122(-5)
0.6	5.230258(-6)	3.329712(-5)	8.853189(-5)	1.724556(-5)	1.097943(-4)	3.149317(-4)	4.194040(-5)
0.7	5.167993(-6)	3.290072(-5)	8.747794(-5)	1.704026(-5)	1.084872(-4)	3.111825(-4)	5.199824(-5)
0.8	5.105728(-6)	3.250433(-5)	8.642399(-5)	1.683495(-5)	1.071801(-4)	3.074333(-4)	3.025500(-5)
0.9	5.043463(-6)	3.210793(-5)	8.537004(-5)	1.662965(-5)	1.058730(-4)	3.036842(-4)	1.541195(-4)
1.0	4.981198(-6)	3.171154(-5)	8.431608(-5)	1.642434(-5)	1.045660(-4)	2.999350(-4)	2.787937(-5)

## Numerical results Graphs for Problem 1



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 $egin{aligned} &y'(t)=-y(t),\ &y(0)=[r-1,1-r], & 0\leq t\leq 1 \end{aligned}$  The equivalent system of ODEs based on (i)-differentiability:

$$(y^{-})'(t;r) = -y^{+}(t;r), \quad y^{-}(0;r) = r-1$$
  
 $(y^{+})'(t;r) = -y^{-}(t;r), \quad y^{+}(0;r) = 1-r$ 

Solutions based on (i)-differentiability:

$$ilde{Y}^{-}(t;r) = (r-1)e^t, \ ilde{Y}^{+}(t;r) = (1-r)e^t$$

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The equivalent system of ODEs based on (ii)-differentiability:

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$$(y^{-})'(t;r) = -y^{-}(t;r), \quad y^{-}(0;r) = r-1$$
  
 $(y^{+})'(t;r) = -y^{+}(t;r), \quad y^{+}(0;r) = 1-r$ 

Solutions based on (ii)-differentiability:

$$ilde{Y}^{-}(t;r) = (r-1)e^{-t}, \ ilde{Y}^{+}(t;r) = (1-r)e^{-t}$$

Source : Nieto et al. (2009)

## Numerical results

#### Numerical results for Problem 2 based on (i)-differentiability

		t = 0.5			t = 1	
	GLM(3)	RK(3)	ABM(3)	GLM(3)	RK(3)	ABM(3)
r	E <sup>-</sup> (0.5; r)	$E^{-}(0.5; r)$	$E^{-}(0.5; r)$	E <sup>-</sup> (1; r)	E <sup>-</sup> (1; r)	$E^{-}(1; r)$
0.0	4.981198(-6)	3.171154(-5)	8.431608(-5)	1.642434(-5)	1.045660(-5)	2.999350(-4)
0.1	4.483078(-6)	2.854039(-5)	7.588448(-5)	1.478191(-5)	9.410938(-5)	2.699415(-4)
0.2	3.984959(-6)	2.536923(-5)	6.745287(-5)	1.313947(-5)	8.365278(-5)	2.399480(-4)
0.3	3.486839(-6)	2.219808(-5)	5.902126(-5)	1.149704(-5)	7.319618(-5)	2.099545(-4)
0.4	2.988719(-6)	1.902692(-5)	5.058965(-5)	9.854606(-6)	6.273959(-5)	1.799610(-4)
0.5	2.490599(-6)	1.585577(-5)	4.215804(-5)	8.212172(-6)	5.228299(-5)	1.499675(-4)
0.6	1.992479(-6)	1.268462(-5)	3.372643(-5)	6.569738(-6)	4.182639(-5)	1.199740(-4)
0.7	1.494359(-6)	9.513462(-6)	2.529482(-5)	4.927303(-6)	3.136979(-5)	8.998049(-5)
0.8	9.962396(-7)	6.342308(-6)	1.686322(-5)	3.284869(-6)	2.091319(-5)	5.998699(-5)
0.9	4.981198(-7)	3.171154(-6)	8.431608(-6)	1.642434(-6)	1.045660(-5)	2.999350(-5)
1.0	0	0	0	0	0	0
r	E <sup>+</sup> (0.5; r)	E <sup>+</sup> (0.5; r)	$E^+(0.5; r)$	$E^+(1; r)$	E <sup>+</sup> (1; r)	E <sup>+</sup> (1; r)
0.0	4.981198(-6)	3.171154(-5)	8.431608(-5)	1.642434(-5)	1.045660(-5)	2.999350(-4)
0.1	4.483078(-6)	2.854039(-5)	7.588448(-5)	1.478191(-5)	9.410938(-5)	2.699415(-4)
0.2	3.984959(-6)	2.536923(-5)	6.745287(-5)	1.313947(-5)	8.365278(-5)	2.399480(-4)
0.3	3.486839(-6)	2.219808(-5)	5.902126(-5)	1.149704(-5)	7.319618(-5)	2.099545(-4)
0.4	2.988719(-6)	1.902692(-5)	5.058965(-5)	9.854606(-6)	6.273959(-5)	1.799610(-4)
0.5	2.490599(-6)	1.585577(-5)	4.215804(-5)	8.212172(-6)	5.228299(-5)	1.499675(-4)
0.6	1.992479(-6)	1.268462(-5)	3.372643(-5)	6.569738(-6)	4.182639(-5)	1.199740(-4)
0.7	1.494359(-6)	9.513462(-6)	2.529482(-5)	4.927303(-6)	3.136979(-5)	8.998049(-5)
0.8	9.962396(-7)	6.342308(-6)	1.686322(-5)	3.284869(-6)	2.091319(-5)	5.998699(-5)
0.9	4.981198(-7)	3.171154(-6)	8.431608(-6)	1.642434(-6)	1.045660(-5)	2.999350(-5)
1.0	0	0	0	0	0	0

## Numerical results

#### Numerical results for Problem 2 based on (ii)-differentiability

		t = 0.5			t = 1	
	GLM(3)	RK(3)	ABM(3)	GLM(3)	RK(3)	ABM(3)
r	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (1; r)	E <sup>-</sup> (1; r)	E <sup>-</sup> (1; r)
0.0	8.520483(-7)	1.369017(-5)	3.852918(-5)	1.033467(-6)	1.660682(-5)	5.046724(-5)
0.1	7.668435(-7)	1.232115(-5)	3.467626(-5)	9.301205(-7)	1.494614(-5)	4.542052(-5)
0.2	6.816386(-7)	1.095213(-5)	3.082334(-5)	8.267738(-7)	1.328546(-5)	4.037379(-5)
0.3	5.964338(-7)	9.583117(-6)	2.697042(-5)	7.234270(-7)	1.162478(-5)	3.532707(-5)
0.4	5.112290(-7)	8.214100(-6)	2.311751(-5)	6.200803(-7)	9.964094(-6)	3.028035(-5)
0.5	4.260241(-7)	6.845083(-6)	1.926459(-5)	5.167336(-7)	8.303412(-6)	2.523362(-5)
0.6	3.408193(-7)	5.476067(-6)	1.541167(-5)	4.133869(-7)	6.642730(-6)	2.018690(-5)
0.7	2.556145(-7)	4.107050(-6)	1.155875(-5)	3.100402(-7)	4.982047(-6)	1.514017(-5)
0.8	1.704097(-7)	2.738033(-6)	7.705835(-6)	2.066934(-7)	3.321365(-6)	1.009345(-5)
0.9	8.520483(-8)	1.369017(-6)	3.852918(-6)	1.033467(-7)	1.660682(-6)	5.046724(-6)
1.0	0	0	0	0	0	0
r	E <sup>+</sup> (0.5; r)	$E^+(0.5; r)$	$E^+(0.5; r)$	$E^+(1; r)$	E <sup>+</sup> (1; r)	E <sup>+</sup> (1; r)
0.0	8.520483(-7)	1.369017(-5)	3.852918(-5)	1.033467(-6)	1.660682(-5)	5.046724(-5)
0.1	7.668435(-7)	1.232115(-5)	3.467626(-5)	9.301205(-7)	1.494614(-5)	4.542052(-5)
0.2	6.816386(-7)	1.095213(-5)	3.082334(-5)	8.267738(-7)	1.328546(-5)	4.037379(-5)
0.3	5.964338(-7)	9.583117(-6)	2.697042(-5)	7.234270(-7)	1.162478(-5)	3.532707(-5)
0.4	5.112290(-7)	8.214100(-6)	2.311751(-5)	6.200803(-7)	9.964094(-6)	3.028035(-5)
0.5	4.260241(-7)	6.845083(-6)	1.926459(-5)	5.167336(-7)	8.303412(-6)	2.523362(-5)
0.6	3.408193(-7)	5.476067(-6)	1.541167(-5)	4.133869(-7)	6.642730(-6)	2.018690(-5)
0.7	2.556145(-7)	4.107050(-6)	1.155875(-5)	3.100402(-7)	4.982047(-6)	1.514017(-5)
0.8	1.704097(-7)	2.738033(-6)	7.705835(-6)	2.066934(-7)	3.321365(-6)	1.009345(-5)
0.9	8.520483(-8)	1.369017(-6)	3.852918(-6)	1.033467(-7)	1.660682(-6)	5.046724(-6)
1.0	0	0	0	0	0	0

## Numerical results Graphs for Problem 2



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### Numerical results Graphs for Problem 2



Figure 2(c) Figure 2(d) Figure 2. (c) 3D-plot with (i)-differentiability; (b) 3D-plot with (ii)-differentiability



• The generalized Hukuhara differetiability was implemented to develop the fuzzy version of a third order GLM and tested on FDEs.



- The generalized Hukuhara differetiability was implemented to develop the fuzzy version of a third order GLM and tested on FDEs.
- From the numerical results, GLM is more accurate than the RK and Adam Bashforth method of same order.

## General Linear Method for Solving Fuzzy Volterra Integro-Differential Equtaions

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The first order fuzzy initial value problem of fuzzy volterra integro-differential equation (FVIDE) of the form

$$y'(t) = f(t,y) + \int_0^x K(t,s)y(s)ds, \quad y(t_0) = y_0 \in \mathbb{R}_F,$$
 (1)

where function  $f : \mathbb{R} \times \mathbb{R}_F \to \mathbb{R}_F$ , crisp function K(t, s) are continuous and  $y_0$  is a fuzzy number.

## P. Linz (1985)

Presented the study of Volterra integro-differential equations.



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Presented the study of Volterra integro-differential equations.

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Presented the methods in finding the numerical results of Volterra integro-differential equations.



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## P. Linz (1985)

Presented the study of Volterra integro-differential equations.

## A. Filiz (2013)

Presented the methods in finding the numerical results of Volterra integro-differential equations.

## M. Matinfar, M. Ghanbari, and M. Nuraei (2013)

Demonstrated the approximation of solutions or fuzzy Volterra-integro differential equations using variational iteration method.



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#### Formulation

From the characterization theorem, the parametric forms of FVIDEs based on (i)-differentiability are as follow:

$$(y_r^-)'(t) = f_r^-(t, y_r^-(t), y_r^+(t)) + \int_0^t K^-(t, s) y(s) ds$$
  
=  $f_r^-(t, y_r^-(t), y_r^+(t), \int_0^t K^-(t, s) y(s) ds),$  (2)

$$(y_r^+)'(t) = f_r^+(t, y_r^-(t), y_r^+(t)) + \int_0^t K^+(t, s) y(s) ds$$
  
=  $f_r^+\left(t, y_r^-(t), y_r^+(t), \int_0^t K^+(t, s) y(s) ds\right), \quad 0 \le s \le t \le 1$  (3)

where

$$egin{aligned} &\mathcal{K}^-(t,s)y(s) = egin{cases} &\mathcal{K}(t,s)y_r^-(s),\mathcal{K}(t,s) \geqslant 0, \ &\mathcal{K}(t,s)y_r^+(s),\mathcal{K}(t,s) \leq 0. \end{aligned} \ &\mathcal{K}^+(s,t)y(s) = egin{cases} &\mathcal{K}(s,t)y_r^+(t),\mathcal{K}(t,s) \geqslant 0, \ &\mathcal{K}(s,t)y_r^-(t),\mathcal{K}(t,s) \leq 0. \end{aligned} \end{aligned}$$



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In the integral part of the equation:  $\int_0^t K(s, t)y(s)ds$ , the integration will be computed using Composite Simpson's Rule. For even sub-intervals:

$$\int_{0}^{t_{n}} \mathcal{K}(t,s) y(s) ds = \mathcal{K}(t,t_{0}) y(t_{0}) + 2 \sum_{j=1}^{\frac{n}{2}-1} \mathcal{K}(t,t_{2m}) y(t_{2m}) + 4 \sum_{j=1}^{\frac{n}{2}} \mathcal{K}(t,t_{2m-1}) y(t_{2m-1}) + \mathcal{K}(t,t_{2n}) y(t_{n})$$
(4)

For odd sub-intervals:

$$\int_{0}^{t_n} K(t,s) y(s) ds = (19) + rac{3}{8} Simpson' srule$$

(5)

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# General Linear Method for Solving FVIDEs Formulation

In the integral part of the equation:  $\int_{t_n}^{t_{n+ch}} K(s,t) x(s) ds$ , on interval  $[t_n, t_{n+\frac{1}{2}h}]$ ,

$$\int_{t_n}^{t_{n+\frac{1}{2}h}} \mathcal{K}(t,s) y(s) ds = h \bigg\{ -\frac{1}{72} \mathcal{K}(t,s) y(t_{n-1}) + \frac{7}{24} \mathcal{K}(t,s) y(t_n) + \frac{2}{9} \mathcal{K}(t,s) y(t_{n+\frac{1}{2}h}) \bigg\},$$
(6)

and on the interval  $[t_n, t_{n+h}]$  we get

$$\int_{t_n}^{t_{n+h}} \mathcal{K}(t,s) y(s) ds = h \bigg\{ -\frac{1}{12} \mathcal{K}(s,t) y(t_{n-1}) + \frac{2}{3} \mathcal{K}(t,s) y(t_n) + \frac{5}{12} \mathcal{K}(t,s) y(t_{n+h}) \bigg\}.$$
(7)

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The fuzzy GLM for solving FVIDEs are similar as equations (7-11), however with extra integration operation as shown below:

$$\begin{split} F_{1}(t_{n}, y(t_{n}; r)) &= \min \left\{ f(t_{n} + c_{1}h, u, v) | u \in [Y_{1}^{-}(y(t_{n}; r)), Y_{1}^{+}(y(t_{n}; r))], v \in [z_{1}^{-}(y(t_{n}; r)), z_{1}^{+}(y(t_{n}; r))] \right\}, \\ G_{1}(t_{n}, y(t_{n}; r)) &= \max \left\{ f(t_{n} + c_{1}h, u, v) | u \in [Y_{1}^{-}(y(t_{n}; r)), Y_{1}^{+}(y(t_{n}; r))] \right\}, v \in [z_{1}^{-}(y(t_{n}; r)), z_{1}^{+}(y(t_{n}; r))] \right\}, \\ F_{2}(t_{n}, y(t_{n}; r)) &= \min \left\{ f(t_{n} + c_{2}h, u, v) | u \in [Y_{2}^{-}(y(t_{n}; r)), Y_{2}^{+}(y(t_{n}; r))] \right\}, v \in [z_{2}^{-}(y(t_{n}; r)), z_{2}^{+}(y(t_{n}; r))] \\ G_{2}(t_{n}, y(t_{n}; r)) &= \max \left\{ f(t_{n} + c_{2}h, u, v) | u \in [Y_{2}^{-}(y(t_{n}; r)), Y_{2}^{+}(y(t_{n}; r))] \right\}, v \in [z_{2}^{-}(y(t_{n}; r)), z_{2}^{+}(y(t_{n}; r))] \\ F_{3}(t_{n}, y(t_{n}; r)) &= \min \left\{ f(t_{n} + c_{3}h, u, v) | u \in [Y_{3}^{-}(y(t_{n}; r)), Y_{3}^{+}(y(t_{n}; r))] \right\}, v \in [z_{3}^{-}(y(t_{n}; r)), z_{3}^{+}(y(t_{n}; r))] \\ G_{3}(t_{n}, y(t_{n}; r)) &= \max \left\{ f(t_{n} + c_{3}h, u, v) | u \in [Y_{3}^{-}(y(t_{n}; r)), Y_{3}^{+}(y(t_{n}; r))] \right\}, v \in [z_{3}^{-}(y(t_{n}; r)), z_{3}^{+}(y(t_{n}; r))] \\ \end{array}$$



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# General Linear Method for Solving FVIDEs

where

$$\begin{aligned} z_1^{-}(y(t_n)) &= \int_0^{t_n} K^{-}(t,s) y(s;r) ds \\ z_1^{+}(y(t_n)) &= \int_0^{t_n} K^{+}(t,s) y(s;r) ds \\ z_2^{-}(y(t_n)) &= \int_0^{t_n} K^{-}(t,s) y(s;r) ds + \int_{t_n}^{t_{n+c_2h}} K^{-}(t,s) y(s) ds \\ z_2^{+}(y(t_n)) &= \int_0^{t_n} K^{+}(t,s) y(s;r) ds + \int_{t_n}^{t_{n+c_2h}} K^{+}(t,s) y(s) ds \\ z_3^{-}(y(t_n)) &= \int_0^{t_n} K^{-}(t,s) y(s) ds + \int_{t_n}^{t_{n+c_3h}} K^{-}(t,s) y(s) ds \\ z_3^{+}(y(t_n)) &= \int_0^{t_n} K^{+}(t,s) y(s) ds + \int_{t_n}^{t_{n+c_3h}} K^{+}(t,s) y(s) ds \end{aligned}$$

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# General Linear Method for Solving FVIDEs Notations

r  $y^{-}$   $\tilde{\gamma}^{+}$   $\tilde{\gamma}^{-}$   $\tilde{\gamma}^{+}$   $E^{-}(t;r)$   $E^{+}(t;r)$  GLM(3) RK(3)

- : r-level set of fuzzy numbers
  - : left bounded of approximate solution
  - : right bounded of approximate solution
  - : left bounded of exact solution
  - : right bounded of exact solution
- : left bounded error of approxiate solution
- : right bounded error of approxiate solution
- : fuzzy third order GLM Set 1
- : fuzzy third order classical Runge-Kutta method from (Butcher, 2008)



# Numerical results Problem 1

$$egin{aligned} y'(t) &= C rac{1}{12t} (36 - 5t^4) + \int_0^t (t^2 + s^2) y(s;r) ds \ C &= [(r^5 + 2r)t^3, (6 - 3r^3)t^3], \quad y(0) = [0,0], \qquad 0 \leq s \leq t \leq 1 \end{aligned}$$

The equivalent system of ODEs based on (i)-differentiability:

$$(y^{-})'(t;r) = \frac{1}{12}rt^{2}(r^{4}+2)(36-5t^{4}) + \int_{0}^{t}(t^{2}+s^{2})y^{-}(s;r), \quad y^{-}(0;r) = 0$$
  
$$(y^{+})'(t;r) = \frac{1}{4}t^{2}(r^{3}-2)(5t^{4}-36) + \int_{0}^{t}(t^{2}+s^{2})y^{+}(s;r), \quad y^{+}(0;r) = 0$$

Solutions :

$$\tilde{Y}^{-}(t;r) = (r^{5} + 2r)t^{3}$$
 $\tilde{Y}^{+}(t;r) = (6 - 3r^{3})t^{3}$ 
Source : Matinfar et al. (2013)

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## Numerical results Problem 1

	t =	0.5	t = 1		
	GLM(3)	RK(3)	GLM(3)	RK(3)	
r	$E^{-}(0.5; r)$	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (1; r)	$E^{-}(1; r)$	
0.0	0	0	0	0	
0.1	9.956810(-11)	4.823713(-10)	1.209550(-9)	7.462263(-9)	
0.2	1.992855(-10)	9.654662(-10)	2.420914(-9)	1.493572(-8)	
0.3	2.998988(-10)	1.452902(-9)	3.643163(-9)	2.247632(-8)	
0.4	4.033500(-10)	1.954086(-9)	4.899886(-9)	3.022960(-8)	
0.5	5.133730(-10)	2.487103(-9)	6.236430(-9)	3.847537(-8)	
0.6	6.360880(-10)	3.081621(-9)	7.727180(-9)	4.767250(-8)	
0.7	7.806100(-10)	3.781770(-9)	9.482820(-8)	5.850381(-8)	
0.8	9.596280(-10)	4.649055(-9)	1.165754(-8)	7.192064(-8)	
0.9	1.190025(-9)	5.765231(-9)	1.445635(-8)	8.918787(-8)	
1.0	1.493445(-9)	7.235207(-9)	1.814234(-8)	1.119283(-7)	
r	$E^+(0.5; r)$	E <sup>+</sup> (0.5; r)	E <sup>+</sup> (1; r)	E <sup>+</sup> (1; r)	
0.0	2.986892(-9)	1.447042(-8)	3.628465(-8)	2.238567(-7)	
0.1	2.985401(-9)	1.446318(-8)	3.626651(-8)	2.237446(-7)	
0.2	2.974945(-9)	1.441254(-8)	3.613950(-8)	2.229612(-7)	
0.3	2.946571(-9)	1.427506(-8)	3.579484(-8)	2.208346(-7)	
0.4	2.891310(-9)	1.400736(-8)	3.512354(-8)	2.166932(-7)	
0.5	2.800214(-9)	1.356601(-8)	3.401690(-8)	2.098656(-7)	
0.6	2.664307(-9)	1.290761(-8)	3.236592(-8)	1.996801(-7)	
0.7	2.474641(-9)	1.198874(-8)	3.006184(-8)	1.854652(-7)	
0.8	2.222250(-9)	1.076599(-8)	2.699583(-8)	1.665493(-7)	
0.9	1.898170(-9)	9.195954(-8)	2.305891(-8)	1.422609(-7)	
1.0	1.493445(-9)	7.235207(-8)	1.814234(-8)	1.119283(-7)	

# Numerical results Graphs for Problem 1



Figure 1. (a) Circle-point: GLM(3), line: Exact; (b) 3D-plot at t = 1.0



## Numerical results Problem 2

$$egin{aligned} y'(t) &= C + \int_0^t (-y(s)) ds, \quad C = [2(r-2) sin(t), 2(2-3r) sin(t)], \ y(0) &= [3r-2,2-r], \qquad 0 \leq t \leq 1 \ \end{aligned}$$
 The equivalent system of ODEs based on (i)-differentiability:

$$(y^{-})'(t;r) = 2(r-2)sin(t) + \int_{0}^{1} (-1)y^{+}(s;r), \quad y^{-}(0;r) = 3r-2$$
  
$$(y^{+})'(t;r) = 2(2-3r)sin(t) + \int_{0}^{1} (-1)y^{-}(s;r), \quad y^{+}(0;r) = 2-r$$

Solutions based on (i)-differentiability:

$$\begin{split} & ilde{Y}^-(t;r) = -rtsin(t) + (2-r)cos(t) + 2(r-1)(exp(t) + exp(-t)) \ & ilde{Y}^+(t;r) = -rtsin(t) + (3r-2)cos(t) + 2(1-r)(exp(t) + exp(-t)) \end{split}$$



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The equivalent system of ODEs based on (ii)-differentiability:

$$(y^{-})'(t;r) = 2(2-3r)sin(t) + \int_{0}^{1} (-1)y^{-}(s;r), \quad y^{-}(0;r) = 3r-2$$
  
 $(y^{+})'(t;r) = 2(r-2)sin(t) + \int_{0}^{1} (-1)y^{+}(s;r), \quad y^{+}(0;r) = 2-r$ 

Solutions based on (ii)-differentiability:

 $ilde{Y}^{-}(t;r) = (3r-2)(cos(t) - tsin(t)), \quad ilde{Y}^{+}(t;r) = (2-r)(cos(t) - tsin(t))$ Source : Ghanbari (2016)

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# Numerical results

#### Numerical results for Problem 2 based on (i)-differentiability

		0.5	. 1	
	t =	0.5	t = 1	
	GLM(3)	RK(3)	GLM(3)	RK(3)
r	$E^{-}(0.5; r)$	E <sup>-</sup> (0.5; r)	E <sup>-</sup> (1; r)	$E^{-}(1; r)$
0.0	9.591260(-9)	3.884529(-8)	1.703350(-8)	8.745231(-8)
0.1	8.252610(-9)	3.333684(-8)	1.499912(-8)	7.665465(-8)
0.2	6.913930(-9)	2.782840(-8)	1.296471(-8)	6.585705(-8)
0.3	5.575350(-9)	2.231994(-8)	1.093031(-8)	5.505935(-8)
0.4	4.236740(-9)	1.681141(-8)	8.895910(-9)	4.426157(-8)
0.5	2.898090(-9)	1.130299(-8)	6.861520(-9)	3.346385(-8)
0.6	1.559467(-9)	5.794547(-9)	4.827130(-9)	2.266616(-8)
0.7	2.208310(-10)	2.860890(-10)	2.792740(-9)	1.186848(-8)
0.8	1.117786(-9)	5.222359(-9)	7.583400(-9)	1.070790(-8)
0.9	2.456414(-9)	1.073082(-8)	1.276027(-9)	9.726932(-8)
1.0	3.795042(-9)	1.623928(-8)	3.310421(-9)	2.052464(-8)
r	E <sup>+</sup> (0.5; r)	E <sup>+</sup> (0.5; r)	$E^+(1; r)$	E <sup>+</sup> (1; r)
0.0	9.591260(-9)	3.884529(-8)	1.703350(-8)	8.745231(-8)
0.1	9.011600(-9)	3.658467(-8)	1.566117(-8)	8.075957(-8)
0.2	8.431990(-9)	3.432415(-8)	1.428888(-8)	7.406693(-8)
0.3	7.852400(-9)	3.206350(-8)	1.291656(-8)	6.737409(-8)
0.4	7.272740(-9)	2.980293(-8)	1.154423(-8)	6.068132(-8)
0.5	6.693130(-9)	2.754225(-8)	1.017195(-8)	5.398850(-8)
0.6	6.113500(-9)	2.528165(-8)	8.799650(-9)	4.729569(-8)
0.7	5.533890(-9)	2.302113(-8)	7.427330(-9)	4.060305(-8)
0.8	4.954250(-9)	2.076049(-8)	6.054989(-9)	3.391018(-8)
0.9	4.374664(-9)	1.849986(-8)	4.682732(-9)	2.721740(-8)
1.0	3.795042(-9)	1.623928(-8)	3.310421(-9)	2.052464(-8)

# Numerical results

#### Numerical results for Problem 2 based on (ii)-differentiability

	t =	0.5	t = 1	
	GLM(3)	RK(3)	GLM(3)	RK(3)
r	E <sup>-</sup> (0.5; r)	$E^{-}(0.5; r)$	E <sup>-</sup> (1; r)	$E^{-}(1; r)$
0.0	7.590100(-9)	3.247856(-8)	6.620856(-9)	4.104928(-8)
0.1	6.451570(-9)	2.760677(-8)	5.627722(-9)	3.489188(-8)
0.2	5.313055(-9)	2.273503(-8)	4.634590(-9)	2.873452(-8)
0.3	4.174528(-9)	1.786324(-8)	3.641453(-9)	2.257713(-8)
0.4	3.036036(-9)	1.299143(-8)	2.648343(-9)	1.641972(-8)
0.5	1.897518(-9)	8.119643(-9)	1.655208(-9)	1.026232(-8)
0.6	7.590100(-10)	3.247856(-9)	6.620856(-10)	4.104928(-9)
0.7	3.795042(-10)	1.623928(-9)	3.310421(-10)	2.052464(-9)
0.8	1.518015(-9)	6.495721(-9)	1.324166(-9)	8.209863(-8)
0.9	2.656530(-9)	1.136750(-8)	2.317297(-9)	1.436725(-8)
1.0	3.795042(-9)	1.623928(-8)	3.310421(-9)	2.052464(-8)
r	$E^+(0.5; r)$	$E^+(0.5; r)$	$E^+(1; r)$	E <sup>+</sup> (1; r)
0.0	7.590100(-9)	3.247856(-8)	6.620856(-9)	4.104928(-8)
0.1	7.210590(-9)	3.085463(-8)	6.289804(-9)	3.899682(-8)
0.2	6.831070(-9)	2.923064(-8)	5.958752(-9)	3.694430(-8)
0.3	6.451570(-9)	2.760677(-8)	5.627722(-9)	3.489188(-8)
0.4	6.072060(-9)	2.598287(-8)	5.296667(-9)	3.283943(-8)
0.5	5.692550(-9)	2.435890(-8)	4.965625(-9)	3.078694(-8)
0.6	5.313055(-9)	2.273503(-8)	4.634590(-9)	2.873452(-8)
0.7	4.933553(-9)	2.111106(-8)	4.303548(-9)	2.668203(-8)
0.8	4.554036(-9)	1.948713(-8)	3.972496(-9)	2.462956(-8)
0.9	4.174528(-9)	1.786324(-8)	3.641453(-9)	2.257713(-8)
1.0	3.795042(-9)	1.623928(-8)	3.310421(-9)	2.052464(-8)



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# Numerical results Graphs for Problem 2



## Numerical results Graphs for Problem 2



Figure 2(c) Figure 2(d) Figure 2. (c) 3D-plot with (i)-differentiability; (b) 3D-plot with (ii)-differentiability



- Simpson's rule and Lagrange interpolating polynomial are explored.
- Based on these integration methods along with the generalized Hukuhara differetiability, the fuzzy version of third order GLM for solving FVIDEs are developed.
- Test results showed that GLM performed better than RK method of same order.

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