

# INTRODUCTION TO NUMERICAL SOLUTION OF FRACTIONAL SYSTEMS

Presented by: Mateo Restrepo S. Juan S. Cárdenas R. David Plazas E. Juan J. Jaramillo C.

Prof.: Samir Posada M.

Numerical Analyisis **EAFIT University** 2019



## OUTLINE

#### 1 INTRODUCTION

- 1.1 Fractional Derivatives
- 1.2 Caputo Definition
- 1.3 Riemann-Liouville Integral
- 1.4 The Tautochrone Problem

#### 2 INTEGER ORDER METHODS

- 2.1 Fourth-Order Runge-Kutta Method (RK4)
- 2.2 Comparison Euler RK4
- 2.3 Systems of ODEs
- 2.4 Multi-Term ODEs

#### 3 FRACTIONAL ORDER METHODS

- 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)
- 3.2 Decomposition Method
- 3.3 Comparison ABM Decomposition
- 3.4 Systems of FDEs
- 3.5 Multi-Term FDEs

#### 4 IMPROVEMENTS

- 4.1 Predicted ABM
- 4.2 Quadrature Decomposition
- 4.3 Polynomial Decomposition
- 4.4 Further Work
- 5 CONCLUSIONS

REFERENCES



#### 1.1 Fractional Derivatives

#### Pros

- Generalization of ordinary derivatives.
- Nonlocal operators Memory and heritage.
- More accuracy and robustness.
- Unexplored areas and applications.

#### Cons

- There multiple fractional are derivatives definitions.
- The derivatives are not always in terms of elementary functions.
- Some definitions require strong conditions on the functions to differentiate.

### 1.2 Caputo Definition

The Caputo definition of fractional derivative is

$$\mathcal{D}_{\mathcal{C}}^{\alpha}y(t) = J^{m-\alpha}y^{(m)}(t) = \frac{1}{\Gamma(m-\alpha)} \int_{0}^{t} \frac{y^{(m)}(\lambda)}{(t-\lambda)^{1-m+\alpha}} d\lambda \tag{1}$$

where  $\alpha > 0$ ,  $m = \lceil \alpha \rceil$ ,  $\Gamma(\cdot)$  is the gamma function and  $J^{m-\alpha}$  is the Riemann-Liouville integral. From now on, the Caputo-type fractional derivative will be denoted as

$$\mathcal{D}_C^{\alpha} y(t) = \frac{d^{\alpha}}{dt^{\alpha}} y(t) \tag{2}$$

For example

$$\begin{split} \frac{d^{0.5}}{dt^{0.5}}[t] &= \frac{1}{\Gamma(1-0.5)} \int_0^t \frac{d}{d\lambda}(\lambda) \cdot \frac{d\lambda}{(t-\lambda)^{1-0.5+1}} \\ &= \frac{1}{\sqrt{\pi}} \int_0^t \frac{d\lambda}{(t-\lambda)^{1/2}} \quad , \text{let } u = t-\lambda \to du = -d\lambda. \\ &= \frac{1}{\sqrt{\pi}} \int_0^t \frac{du}{u^{1/2}} \\ &= 2\sqrt{\frac{t}{\pi}} \end{split}$$

### 1.3 Riemann-Liouville Integral

The Riemann-Liouville integral of order  $\alpha$  is defined as

$$J^{\alpha}y(t) = \frac{1}{\Gamma(\alpha)} \int_{0}^{t} \frac{y(\lambda)}{(t-\lambda)^{1-\alpha}} d\lambda$$
 (3)

### **Properties**

$$J^{\alpha}\left[f(t) + g(t)\right] = J^{\alpha}f(t) + J^{\alpha}g(t) \tag{4}$$

$$J^{\alpha}J^{\beta}f(t) = J^{\alpha+\beta}f(t) \tag{5}$$

$$\frac{d^{\alpha}}{dt^{\alpha}} \left[ J^{\alpha} y(t) \right] = y(t) \tag{6}$$

$$J^{\alpha} \left[ \frac{d^{\alpha}}{dt^{\alpha}} y(t) \right] = y(t) - \sum_{r=0}^{m-1} \frac{y_r t^r}{r!}$$
 (7)

#### 1.4 The Tautochrone Problem

It is desired to find a curve such that, if an object starts on any point along this curve, the time that it requires to slide down to the origin is the same.

- Tauto  $\rightarrow$  equal.
- Chrono  $\rightarrow$  time.
- Abel, XVIII century.

### Assumptions:

- The object moves only by the force of gravity.
- It moves without friction.

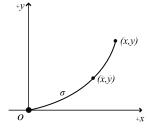


Figure 1. Tautochrone problem.

Click for GIF

### 1.4 The Tautochrone Problem

Using the energy conservation law,

$$mgy = \frac{1}{2}m\left(\frac{d\sigma}{dt}\right)^{2} + mg\hat{y}$$

$$\vdots$$

$$\frac{d^{0.5}}{dy^{0.5}}\sigma(y) = \frac{\sqrt{2g}}{\Gamma\left(\frac{1}{2}\right)}T$$

Through some analytical procedures, we obtain

$$x = \frac{gT^2}{\pi^2}[t + \sin(t)]$$
$$y = \frac{gT^2}{\pi^2}[1 - \cos(t)]$$

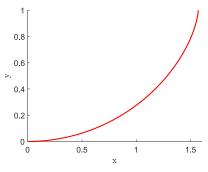


Figure 2. Tautochrone curve.

The following initial value problem (IVP) will be treated:

$$\begin{cases} y' = f(t, y) \\ y(0) = y_0, & t \in [0, T] \end{cases}$$
 (8)

Example:

$$\begin{cases} y' - y = 0 \\ y(0) = -3 \end{cases} \tag{9}$$

$$y(t) = -3e^{-t} \tag{10}$$

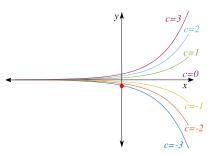


Figure 3. Solution for problem 9.

# 2.1 Fourth-Order Runge-Kutta Method (RK4)

- Improved Euler with more accuracy.
- $4^{th}$  order  $\rightarrow$  optimal.

This method approximates the solution to the IVP as follows:

$$y_{i+1} = y_i + h\left(\frac{k_1 + 2k_2 + 2k_3 + k_4}{6}\right)$$
 (11)

where

$$k_1 = f(t_i, y_i)$$

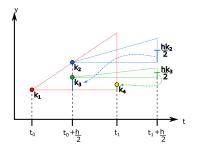
$$k_2 = f(t_i + h/2, y_i + hk_1/2)$$

$$k_3 = f(t_i + h/2, y_i + hk_2/2)$$

$$k_4 = f(t_i + h, y_i + hk_3)$$

The algorithm works as

$$y = runge_kutta(f, y0, T, N)$$

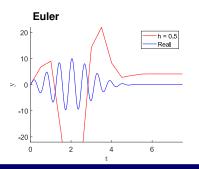


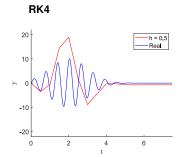
### 2.2 Comparison Euler - RK4

Example: approximate a solution to the IVP

$$\begin{cases} y' = 10e^{-\frac{(t-2)^2}{2}} \left(10\cos(10t) - (t-2)\sin(10t)\right) \\ y(0) = 0 \end{cases}$$
 (12)

$$y(t) = 10e^{-\frac{(t-2)^2}{2}}\sin(10t)$$
 (13)



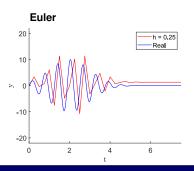


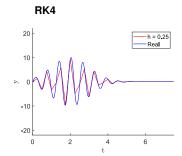
### 2.2 Comparison Euler - RK4

Example: approximate a solution to the IVP

$$\begin{cases} y' = 10e^{-\frac{(t-2)^2}{2}} \left(10\cos(10t) - (t-2)\sin(10t)\right) \\ y(0) = 0 \end{cases}$$
 (12)

$$y(t) = 10e^{-\frac{(t-2)^2}{2}}\sin(10t)$$
 (13)



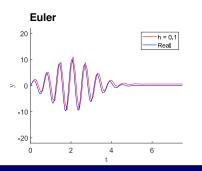


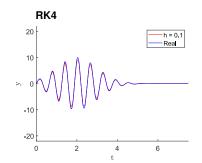
### 2.2 Comparison Euler - RK4

Example: approximate a solution to the IVP

$$\begin{cases} y' = 10e^{-\frac{(t-2)^2}{2}} \left(10\cos(10t) - (t-2)\sin(10t)\right) \\ y(0) = 0 \end{cases}$$
 (12)

$$y(t) = 10e^{-\frac{(t-2)^2}{2}}\sin(10t)$$
 (13)



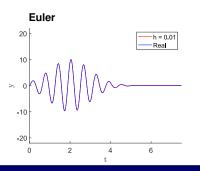


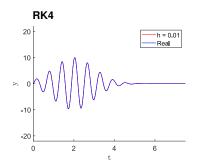
### 2.2 Comparison Euler - RK4

Example: approximate a solution to the IVP

$$\begin{cases} y' = 10e^{-\frac{(t-2)^2}{2}} \left(10\cos(10t) - (t-2)\sin(10t)\right) \\ y(0) = 0 \end{cases}$$
 (12)

$$y(t) = 10e^{-\frac{(t-2)^2}{2}}\sin(10t)$$
 (13)





### 2.3 Systems of ODEs

Consider the system of ordinary differential equations

$$\begin{cases} y'_1 = f_1(t, y_1, y_2, \dots, y_n) & y_1(0) = y_1 \\ y'_2 = f_2(t, y_1, y_2, \dots, y_n) & y_2(0) = y_2 \\ \vdots \\ y'_n = f_n(t, y_1, y_2, \dots, y_n) & y_n(0) = y_n \end{cases}$$
(14)

which can be synthesized as

$$\begin{cases} \mathbf{y}' = F(t, \mathbf{y}) \\ \mathbf{y}(0) = \mathbf{y}_0 \end{cases} \tag{15}$$

For example, the Lotka-Volterra equations (predator-prey model) is a system of ODEs as follows:

$$\begin{cases} y_1' = \alpha y_1 - \beta y_1 y_2 \\ y_2' = \delta y_1 y_2 - \gamma y_2 \end{cases}$$
 (16)

where  $y_1$  represents the number of preys and  $y_2$  is the amount of predators.



### 2.3 Systems of ODEs

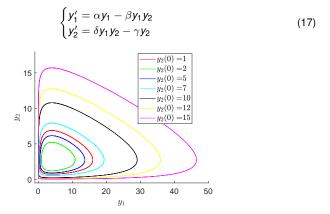


Figure 4. Phase portrait of Lotka-Volterra equations.

#### 2.4 Multi-Term ODEs

In case of higher order ODEs, they can be transformed into a sytem of first order ODEs, using phase variables  $x_i$ . Suppose we have an equation as the following:

$$\begin{cases} y^{(n)} = f(t, y, y', ..., y^{(n-1)}) \\ y(0) = y_0, ..., y^{(n-1)}(0) = y_{(n-1)}, \ t \in [0, T] \end{cases}$$
(18)

with the following substitution

$$\begin{cases} x_{1} = y \\ x_{2} = y' \\ \vdots \\ x_{n} = y^{(n-1)} \end{cases} \longrightarrow \begin{cases} x'_{1} = x_{2} \\ x'_{2} = x_{3} \\ \vdots \\ x'_{n-1} = x_{n} \\ x'_{n} = f(t, x_{1}, x_{2}, ..., x_{n-1}, x_{n}) \end{cases}$$
(19)

$$x_1(0) = y_0, \ldots, x_{n-1}(0) = y_{(n-1)}$$

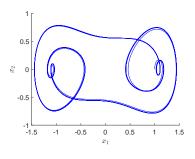
### 2.4 Multi-Term ODEs

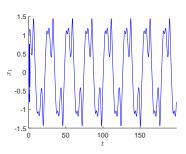
Example: consider the Duffing oscillator (Click for GIF)

$$y'' + \delta y' + \alpha y + \beta y^3 = \gamma \cos(\omega t)$$
 (20)

the equivalent system of first order ODEs is

$$\begin{cases} x_1 = y \\ x_2 = y' \end{cases} \longrightarrow \begin{cases} x_1' = x_2 \\ x_2' = -\delta x_2 - \alpha x_1 - \beta x_1^3 + \gamma \cos(\omega t) \end{cases}$$
 (21)





$$\delta = 0.3, \, \alpha = -1, \, \beta = 1, \gamma = 0.37, \, \omega = 1.2$$
 and initial conditions  $x_1(0) = 1, \, x_2(0) = 0$ .

Consider the fractional IVP

$$\begin{cases} \frac{d^{\alpha}}{dt^{\alpha}} y(t) = f(t, y) \\ y(0) = y_0, \dots, y^{(m-1)}(0) = y_{(m-1)} & \alpha \in \mathbb{R}^+ \quad t \in [0, T] \end{cases}$$
 (22)

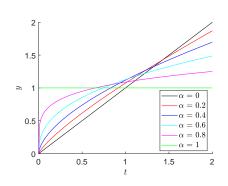
where  $m = \lceil \alpha \rceil$ .

### Example

$$\frac{d^{0.5}}{dt^{0.5}}y(t) = 2\sqrt{\frac{y}{\pi}}$$
 (23)

whose solution is

$$y(t) = t \tag{24}$$



## 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)

Using quadrature theory, the solution can be approximated as

$$y_{h}(t_{n+1}) = \sum_{k=0}^{\lfloor \alpha \rfloor - 1} \frac{t_{n+1}^{k}}{k!} y_{0}^{(k)} + \frac{h^{\alpha}}{\Gamma(\alpha + 2)} f(t_{n+1}, y_{h}^{p}(t_{n+1})) + \frac{h^{\alpha}}{\Gamma(\alpha + 2)} \sum_{j=0}^{n} a_{j,n+1} f(t_{j}, y_{h}(t_{j}))$$
(25)

- $\triangleright$   $y_h^p(t_{n+1})$  is a predicted value.
- $ightharpoonup a_{i,n+1}$  is a quadrature coefficient.

The algorithm works as

$$y = abm(f, alpha, y0, T, N)$$

- f is the right-hand side of the differential equation.
- alpha is the order of the differential equation.
- v0 is the initial conditions.
- T is the simulation time.
- N is the number of partitions on the interval [0, T].



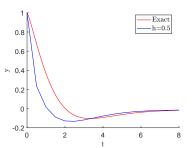
### 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (26)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(27)



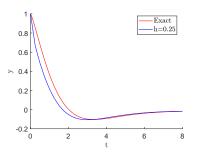
### 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (26)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(27)



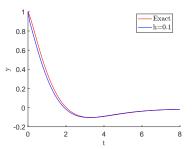
### 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (26)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(27)



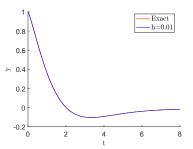
### 3.1 Adams-Bashforth-Moulton Predictor-Corrector (ABM)

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (26)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(27)



### 3.2 Decomposition Method

Based on decomposing f as follows

$$f(t,\mathbf{y}) = g(t) + \mathbf{A}\mathbf{y} + h(t,\mathbf{y}) \tag{28}$$

Applying the inverse operation to the Caputo fractional derivative

$$\mathbf{y}(t) = \sum_{r=0}^{m-1} \frac{\mathbf{y}_r t^r}{r!} + J^{\alpha} g(t) + J^{\alpha} \mathbf{A} \mathbf{y} + J^{\alpha} h(t, \mathbf{y})$$
 (29)

and supposing a solution in series, we obtain the recursive scheme

$$\mathbf{x}_{0} = \sum_{r=0}^{m-1} \frac{\mathbf{y}_{r} t^{r}}{r!} + J^{\alpha} g(t)$$

$$\mathbf{x}_{k+1} = J^{\alpha} \mathbf{A} \mathbf{x}_{k} + J^{\alpha} \tilde{h}_{k} \left( t, \sum_{r=0}^{k} \mathbf{x}_{j}(t) \right)$$
(30)

Where  $\tilde{h}_k$  is the Adomian polynomial

$$\tilde{h}_k\left(t, \sum_{r=0}^k \mathbf{x}_j(t)\right) = \frac{1}{k!} \left[ \frac{d^k}{d\lambda^k} h\left(t, \sum_{j=0}^k \lambda^j \mathbf{x}_j(t)\right) \bigg|_{\lambda=0} \right]$$
(31)

### 3.2 Decomposition Method

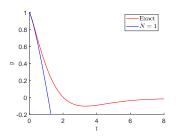
The algorithm works as

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (32)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(33)



### 3.2 Decomposition Method

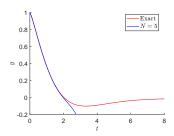
The algorithm works as

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (32)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(33)



### 3.2 Decomposition Method

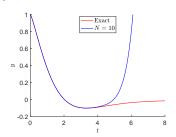
The algorithm works as

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (32)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(33)



### 3.2 Decomposition Method

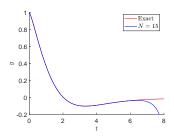
The algorithm works as

### Example

Give an approximate solution to

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
 (32)

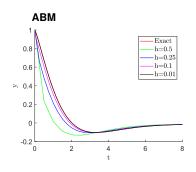
$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(33)

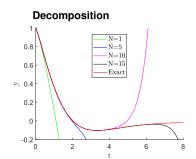


### 3.3 Comparison ABM - Decomposition

$$\begin{cases} \frac{d^{1.25}}{dt^{1.25}}y(t) = -y(t) \\ y'(0) = 0, \ y(0) = 1 \end{cases}$$
(34)

$$y(t) = E_{1.25, 1}(-t^{1.25}) = \sum_{k=0}^{\infty} \frac{(-1)^k t^{1.25k}}{\Gamma(1.25k+1)}$$
(35)





### 3.4 Systems of FDEs

Consider the system of ordinary fractional differential equations

$$\begin{cases} \frac{d^{\alpha_{1}}}{dt^{\alpha_{1}}}y_{1} = f_{1}(t, y_{1}, y_{2}, \dots, y_{n}) & y_{1}(0) = y_{1} \\ \frac{d^{\alpha_{2}}}{dt^{\alpha_{2}}}y_{2} = f_{2}(t, y_{1}, y_{2}, \dots, y_{n}) & y_{2}(0) = y_{2} \\ \vdots & & \vdots \\ \frac{d^{\alpha_{n}}}{dt^{\alpha_{n}}}y_{n} = f_{n}(t, y_{1}, y_{2}, \dots, y_{n}) & y_{n}(0) = y_{n} \\ \alpha_{j} \in \mathbb{R}^{+}, \ t \in [0, T] \end{cases}$$
(36)

which can be synthesized as

$$\begin{cases}
\frac{d^{\alpha}}{dt^{\alpha}}\mathbf{y} = F(t, \mathbf{y}) \\
\mathbf{y}(0) = \mathbf{y}_{0}, \ \alpha \in (\mathbb{R}^{+})^{n \times 1}, \ t \in [0, T]
\end{cases}$$
(37)

#### 3.5 Multi-Term FDEs

Suppose we have the multi-term fractional differential equation

$$\begin{cases}
\frac{d^{\alpha_n}}{dt^{\alpha_n}}y(t) = f\left(t, y, \frac{d^{\alpha_1}}{dt^{\alpha_1}}y, \frac{d^{\alpha_2}}{dt^{\alpha_2}}y, \dots, \frac{d^{\alpha_n}}{dt^{\alpha_n}}y\right) \\
y(0) = y_0, \dots, y^{(m-1)}(0) = y_{(m-1)}
\end{cases}$$
(38)

where  $m = [\alpha_n]$  and  $0 < \alpha_1 < \alpha_2 < \ldots < \alpha_n$ . We select new orders  $\tilde{\alpha}_1, \ldots, \tilde{\alpha}_n$  such that

- (a)  $\tilde{\alpha}_1, \ldots, \tilde{\alpha}_n$  must be rational numbers.
- (b)  $\lceil \alpha_n \rceil = \lceil \tilde{\alpha}_n \rceil$
- (c)  $gcd(1, \tilde{\alpha}_1, \dots, \tilde{\alpha}_n)$  should be as large as possible,
- (d)  $|\alpha_i \tilde{\alpha}_i|$  should be as small as possible for all j

#### 3.5 Multi-Term FDEs

We build the approximated system of FDEs with

$$\gamma := \gcd\left(1, \tilde{\alpha}_{1}, \dots, \tilde{\alpha}_{n}\right)$$

$$\tilde{N} := \frac{\tilde{\alpha}_{n}}{\gamma}$$

$$\begin{cases} \frac{d^{\gamma}}{dt^{\gamma}} x_{0} = x_{1} \\ \frac{d^{\gamma}}{dt^{\gamma}} x_{1} = x_{2} \\ \vdots \\ \frac{d^{\gamma}}{dt^{\gamma}} x_{\tilde{N}-2} = x_{\tilde{N}-1} \\ \frac{d^{\gamma}}{dt^{\gamma}} x_{\tilde{N}-1} = f\left(t, x_{0}, x_{\tilde{\alpha}_{1}/\gamma}, \dots, x_{\tilde{\alpha}_{n-1}/\gamma}\right) \end{cases}$$

$$x_{j}(0) = \begin{cases} y_{(j\gamma)} & \text{for } j\gamma \in \mathbb{N}_{0} \\ 0 & \text{else} \end{cases}$$

$$(40)$$
Create Transform | Vigilada Mineducación

#### 3.5 Multi-Term FDEs

Example: Bagley-Torvik Equation (Click for GIF)

$$\begin{cases} ay'' + b\frac{d^{3/2}}{dt^{3/2}}y + cy = g(t) \\ y(0) = y_0, \ y'(0) = y_1 \end{cases}$$
(41)

Let us keep the original orders  $\tilde{\alpha}_1=3/2$  and  $\tilde{\alpha}_2=2$  to satisfy condition (d). Note that  $\gamma=\gcd(1,3/2,2)=1/2.$  Then  $\tilde{N}=\frac{\dot{\tilde{\alpha}}_2}{\gamma}=4.$  Therefore, the approximated system is

$$\begin{cases} \frac{d^{1/2}}{dt^{1/2}} x_0 = x_1 & x_0(0) = y_0 \\ \frac{d^{1/2}}{dt^{1/2}} x_1 = x_2 & x_1(0) = 0 \\ \frac{d^{1/2}}{dt^{1/2}} x_2 = x_3 & x_2(0) = y_1 \\ \frac{d^{1/2}}{dt^{1/2}} x_3 = f(t, x_0, x_{1.5/0.5}) = \frac{g(t) - cx_0 - bx_3}{a} & x_3(0) = 0 \end{cases}$$

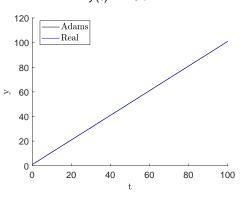
$$(42)$$

### 3.5 Multi-Term FDEs

**Example: Bagley-Torvik Equation** 

In particular, for a = 1, b = c = -1, g(t) = t + 1,  $y_0 = 1$  and  $y_1 = 1$ , the exact solution to this IVP is

$$y(t) = 1 + t \tag{43}$$



### 4.1 Predicted ABM

- + accuracy  $\implies$  + execution time.
- ABM converges  $\iff$  Predicted ABM converges.

The idea is to calculate different  $y_h^p(t_{n+1})$  until a tolerance is reached, for each instant of time.

$$y = pabm(f, alpha, y0, T, N, nmax, tol)$$

For example, using the same Bagley-Torvik equation with larger time step, we have

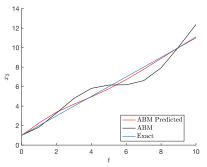


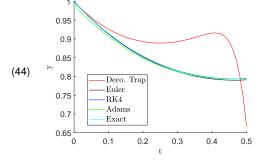
Figure 5. Comparison between the original and predicted scheme.

### 4.2 Quadrature Decomposition

The idea was to partition the time interval and, on each point, find the polynomial using approximations to the Riemann-Liouville integral.

### Example:

$$\begin{cases} x' = y & x(0) = 1 \\ y' = 2x - y & y(0) = -1 \end{cases}$$



### 4.3 Polynomial Decomposition

- Analytic solution.
- execution time.
- Useful for non-chaotic dynamic systems or chaotic systems for small time frames.

The main idea is based on the simple computation of

$$J^{\alpha}\left(t^{\beta}\right) = \frac{\Gamma(\beta+1)}{\Gamma(\alpha+\beta+1)}t^{\alpha+\beta} \tag{45}$$

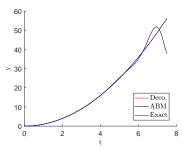
which can be extended to non-polynomial expressions using interpolation.

### 4.3 Polynomial Decomposition

Example:

$$\begin{cases} \frac{d^3}{dt^3}y(t) + \frac{d^{5/2}}{dt^{5/2}}y(t) + y^2(t) = t^4\\ y(0) = y'(0) = 0, \ y''(0) = 2 \end{cases}$$
(46)

The exact solution is  $y(t) = t^2$ . Using the procedure for multi-term FDEs, the solution can be approximated as shown in the figure.

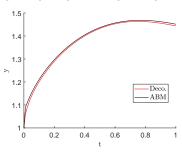


### 4.3 Polynomial Decomposition

**Example: Fractional Lotka-Volterra model** 

$$\begin{cases} \frac{d^{\alpha_1}}{dt^{\alpha_1}} x = \alpha x - \beta xy \\ \frac{d^{\alpha_2}}{dt^{\alpha_2}} y = \delta xy - \gamma y \end{cases}$$
(47)

Using params= [1, 1, 1, 1], c.i= [1, 0.5] and  $\alpha$  = [0.5, 0.6]:

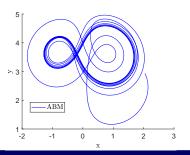


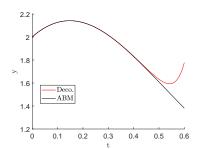
### 4.3 Polynomial Decomposition

**Example: Financial System** 

$$\begin{cases} \frac{d^{\alpha_1}}{dt^{\alpha_1}}x = z + (y - a)x \\ \frac{d^{\alpha_2}}{dt^{\alpha_2}}y = 1 - by - x^2 \\ \frac{d^{\alpha_3}}{dt^{\alpha_3}}z = -x - cz \end{cases}$$
(48)

Using params= [3, 0.1, 1], c.i= [2, 3, 2] and  $\alpha = [1, 1, 0.8]$ , we obtain





### 4.4 Further Work

ABM with fixed memory.

$$\frac{1}{\Gamma(1-\alpha)} \int_{t-T}^{t} \frac{y'(s)}{(t-s)^{\alpha}} ds$$

ABM with logarithmic memory.

$$w^{p\alpha} \int_0^t \frac{f(w^p x)}{(t-x)^{1-\alpha}} dx$$

Richardson extrapolation.

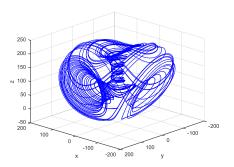
$$x_n = x(t_n) + \sum_{\mu=1}^{M_1} \gamma_{\mu} n^{-\lambda \mu}$$

Decomposition acceleration.

$$S_n^{(k)} = \frac{S_n^{(k-1)} S_{n+2}^{(k-1)} - \left(S_{n+1}^{(k-1)}\right)^2}{S_n^{(k-1)} + S_{n+2}^{(k-1)} - 2S_{n+1}^{(k-1)}}, \quad k \ge 1$$

### 5. CONCLUSIONS

- Fractional calculus ---> powerful tool to model real and chaotic systems.
- Both ABM and decomposition are useful but each with pros and cons.
- Both methods can be improved, as shown.
- A summary of some essential tools to study fractional systems was successfully constructed.



### REFERENCES I

- W. Deng and C. Li. "Numerical schemes for fractional ordinary differential equations." in Numerical modelling. IntechOpen, 2012.
- ▶ R. Almeida, N. R. Bastos, and M. T. T. Monteiro, "Modeling some real phenomena by fractional differential equations," Mathematical Methods in the Applied Sciences, vol. 39, no. 16, pp. 4846-4855, 2016,
- ▶ P. J. Antsaklis and A. N. Michel, *Linear systems*. Springer Science & Business Media, 2006. p. 13.
- ▶ K. Diethelm and J. Ford, "Numerical solution of the bagley-torvik equation," BIT Numerical Mathematics, vol. 42, no. 3, pp. 490-507, 2002.
- ► R. Thomas, "Deterministic chaos seen in terms of feedback circuits: Analysis, synthesis," labyrinth chaos"." International Journal of Bifurcation and Chaos, vol. 9, no. 10, pp. 1889-1905, 1999.
- S. Momani and Z. Odibat. "Numerical approach to differential equations of fractional order." Journal of Computational and Applied Mathematics, vol. 207, no. 1, pp. 96-110, 2007.
- ▶ W. F. Langford, "Numerical studies of torus bifurcations," in *Numerical Methods for Bifurcation* Problems. Springer, 1984, pp. 285–295.
- ▶ W.-C. Chen, "Nonlinear dynamics and chaos in a fractional-order financial system," Chaos, Solitons & Fractals, vol. 36, no. 5, pp. 1305-1314, 2008.

### REFERENCES II

- ► M. Ishteva, "Properties and applications of the caputo fractional operator," Department of Mathematics. University of Karlsruhe. Karlsruhe. 2005.
- ► C. Yang, W. Xiang, and Q. Ji, "Generation of fractional-order chua's chaotic system and it's synchronization." in 2018 Chinese Control And Decision Conference (CCDC). IEEE, 2018, pp. 599-603.
- ▶ D. Rowell, "State-space representation of Iti systems," MIT, 2002.
- ▶ J. C. Sprott and K. E. Chlouverakis, "Labyrinth chaos," International Journal of Bifurcation and Chaos. vol. 17. no. 06. pp. 2097–2108. 2007.
- ▶ U. E. Kocamaz, A. Göksu, H. Taskın, and Y. Uvaroğlu, "Synchronization of chaos in nonlinear finance system by means of sliding mode and passive control methods: a comparative study." Information Technology and Control, vol. 44, no. 2, pp. 172-181, 2015.
- ► P. Dawkins. Euler's method. http://tutorial.math.lamar.edu/Classes/DE/EulersMethod.aspx.
- ▶ O. García Jaimes, J. A. Villegas Gutiérrez, J. I. Castaño Bedoya, and J. A. Sánchez Cano, Ecuaciones Diferenciales. Fondo Editorial Universidad EAFIT. 2016.
- ▶ J. Mathews and K. Fink, Numerical Methods Using MATLAB, ser. Featured Titles for Numerical Analysis Series. Pearson Prentice Hall. 2004. https://books.google.com.co/books? id=E5IRAAAAMAAJ.

### REFERENCES III

- ► T. Kisela, "Fractional differential equations and their applications," Faculty Of Mechanical Engineering. Institute Of Mathemathics, 2008.
- ► K. Diethelm, N. J. Ford, and A. D. Freed, "A predictor-corrector approach for the numerical solution of fractional differential equations," Nonlinear Dynamics, vol. 29, no. 1-4, pp. 3-22, 2002.
- ▶ P. J. Torvik and R. L. Bagley, "On the appearance of the fractional derivative in the behavior of real materials," Journal of Applied Mechanics, vol. 51, no. 2, pp. 294-298, 1984.

