

Two new numerical approaches for the fractional distribution of the model of a system of lakes via modified hat and quasi-hat functions

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Abstract

In this article, two numerical approaches are presented to solve a system of three fractional differential equations that express the pollution of lakes. In our recent study, a new class of hat functions, called quasi-hat functions (QHF), are constructed. The proposed approaches utilize modified hat functions (MHF) and quasi-hat functions (QHF). Fractional-order operational of MHF and QHF are used to build algorithms that transform the main problem into a system of six equations with six unknowns and three equations with three unknowns, respectively. Absolute errors of obtained approximate solutions and convergence analysis of the utilized approach will be studied. Finally, three examples are provided to illustrate the capabilities of these algorithms. The pollution monitoring results are reported in some tables and figures for different values of α .

Keywords: Numerical algorithms; Fractional modeling; Pollution; Systems of lakes; Fractional operational matrix.

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1. Introduction

Water pollution causes very destructive effects on the environment. Researchers use mathematical modeling to monitor pollution and help plan ways to protect the environment. Biazar et al. studied and solved the mathematical model of lake pollution for the first time [1]. This model consists of a system of three lakes connected to each other by some channels. Mathematical modeling of this problem leads to a system of three fractional differential equations. There are several numerical methods for solving such systems of differential equations (see [2, 3, 4, 5]). The model has been examined by many researchers using different approaches [6, 7, 8]. Khader et al. used the operational matrix method based on the shifted Chebyshev polynomials to solve the water pollution model [9]. Prakasha and Veeresha inspected the model of pollution for a system of lakes using the q-homotopy analysis transform method [10]. The Haar wavelet collocation method has been suggested for solving a novel model for the contamination of a system of three artificial lakes in [11]. Ghosh et al. established a new iterative method to solve the model of the amount of pollution in lakes connected with some rivers [12]. Shiri et al. presented a high-order numerical method to solve the pollution model [13]. Also, Yönet et al. applied a Taylor series to solve the pollution system [14]. Figure 1 shows the three lakes with interconnected canals. A pollutant enters the first lake from the indicated source, named by $p(t)$. Suppose $u_i(t)$ and V_i express the amount of pollutant and the water volume in the lake i for $t \geq 0$, $i = 1, 2, 3$, respectively. Modeling the dynamic behavior of pollution distribution in a system of lakes can be stated as follows:

$$\begin{aligned} {}_0^C D_t^\alpha u_1(t) &= p(t) + \frac{F_{13}}{V_3} u_3(t) - \frac{F_{31}}{V_1} u_1(t) - \frac{F_{21}}{V_1} u_1(t), \\ {}_0^C D_t^\alpha u_2(t) &= \frac{F_{21}}{V_1} u_1(t) - \frac{F_{32}}{V_2} u_2(t), \\ {}_0^C D_t^\alpha u_3(t) &= \frac{F_{31}}{V_1} u_1(t) + \frac{F_{32}}{V_2} u_2(t) - \frac{F_{13}}{V_3} u_3(t), \end{aligned} \quad 0 < \alpha \leq 1, \quad t \in [0, T], \quad (1)$$

subject to the initial conditions $u_1(0) = \lambda_1$, $u_2(0) = \lambda_2$, and $u_3(0) = \lambda_3$. Here $u_1(t)$, $u_2(t)$, and $u_3(t)$ are unknown functions, λ_1 , λ_2 , λ_3 , F_{13} , F_{31} , F_{21} , F_{32} ,

V_1 , V_2 , and V_3 are the appropriate parameters. The operator ${}_0^C D_t^\alpha$ denotes the Caputo fractional derivative [15]. Table 1 shows the meaning of parameters and variables for the pollution problem in the system of lakes. The volume of water in each lake is assumed to be constant and the flow into each lake must balance the outflow. The present study discusses some of the properties of Riemann-Liouville integral operators based on the modified hat and quasi-hat functions to solve a fractional distribution model of lakes (1) for the first time. In Section 2, some characteristics and basic definitions of the fractional calculus are explained. Section 3 is dedicated to introducing the operational matrices of MHFs and QHFs. Fourth section studies the absolute error of approximation of a function by a truncated series of MHFs and QHFs. Fifth section presents two numerical algorithms for problem (1). The principal problem will be reduced to several simple linear algebraic equations by applying the operational matrix methods of MHFs and QHFs. The convergence analysis of the proposed schemes is discussed in Section 6. As evidence of the validity and accuracy of the utilized approach, three numerical examples are provided in Section 7, and a conclusion and discussion are provided in Section 8.

2. Basic concepts and definitions

Here, several definitions and properties are explained that will be used in this manuscript. In this research the Riemann-Liouville integral operator of the α -th order (I_t^α), and the Caputo fractional differential operator of order α (${}_0^C D_t^\alpha$) will be used. They are well addressed in [15]. The Riemann-Liouville integral and the Caputo fractional derivative operators satisfies the following properties:

$$\begin{aligned}
 I_t^\alpha(I_t^\beta u(t)) &= I_t^\beta(I_t^\alpha u(t)) = I_t^{\alpha+\beta} u(t), \\
 I_t^\alpha({}_0^C D_t^\alpha u(t)) &= u(t) - \sum_{i=0}^{n-1} u^{(i)}(0) \frac{t^i}{i!}, \quad n-1 < \alpha \leq n, \quad t > 0.
 \end{aligned} \tag{2}$$

2.1. Recalling of MHFs

Hat functions are defined on a closed interval $[0, T]$ and have shapes similar to hats[16, 17]. The interval is segregated into n number of sub-intervals of

equal length, as $[ih, (i+1)h]$, $i = 0, 1, 2, \dots, n-1$, where $h = \frac{T}{n}$, and $n \geq 2$, $n = 2K$, $K \in \mathbb{N}$. Modified hat functions are defined as follows [16]:

$$\psi_0(t) = \begin{cases} \frac{1}{2h^2}(t-h)(t-2h), & 0 \leq t \leq 2h, \\ 0, & \textit{otherwise}, \end{cases} \quad (3)$$

55 if i is odd, and $1 \leq i \leq n-1$;

$$\psi_i(t) = \begin{cases} \frac{-1}{h^2}(t-(i-1)h)(t-(i+1)h), & (i-1)h \leq t \leq (i+1)h, \\ 0, & \textit{otherwise}, \end{cases} \quad (4)$$

if i is even, and $2 \leq i \leq n-2$;

$$\psi_i(t) = \begin{cases} \frac{1}{2h^2}(t-(i-1)h)(t-(i-2)h), & (i-2)h \leq t \leq ih, \\ \frac{1}{2h^2}(t-(i+1)h)(t-(i+2)h), & ih \leq t \leq (i+2)h, \\ 0, & \textit{otherwise}, \end{cases} \quad (5)$$

and at the last point

$$\psi_n(t) = \begin{cases} \frac{1}{2h^2}(t-(T-h))(t-(T-2h)), & T-2h \leq t \leq T, \\ 0, & \textit{otherwise}. \end{cases} \quad (6)$$

2.2. Definition of QHFs

The concept of quasi-hat functions (QHFs) on a closed interval $[0, T]$ is
60 derived based on the idea of the hat functions [18]. The domain of the quasi-hat functions is the same as those of the hat functions that were introduced right now. Quasi-hat functions are defined as follows for i even, and $0 \leq i \leq n$;

$$\phi_i(t) = \begin{cases} \frac{1}{2h^2}(t-(i+1)h)(t-(i+2)h), & ih \leq t < (i+2)h, \\ 0, & \textit{otherwise}, \end{cases} \quad (7)$$

when i is odd, and $1 \leq i \leq n-1$;

$$\phi_i(t) = \begin{cases} -\frac{1}{2h^2}(t-(i-1)h)(t-(i+2)h), & (i-1)h \leq t < (i+1)h, \\ 0, & \textit{otherwise}, \end{cases} \quad (8)$$

wherein $n \geq 2$ is an even positive integer, $h = \frac{T}{n}$. The Matlab package is used
65 to plot quasi-hat functions on the interval $[0, 1]$ for $n = 4$ (Figure 2).

The following properties can be achieved by using the MHFs and QHFs definitions, respectively [16, 18];

$$\psi_i(jh) = \begin{cases} 1, & i = j, \\ 0, & i \neq j, \end{cases} \quad \sum_{i=0}^n \psi_i(t) = 1, \quad (9)$$

and

$$\phi_i(jh) = \begin{cases} 1, & i = j, \\ 0, & i \neq j, \end{cases} \quad \sum_{i=0}^n \phi_i(t) = 1. \quad (10)$$

2.3. MHFs & QHFs expansion

70 An arbitrary function $u(t)$, can be approximated by a linear combination of MHFs or QHFs as the following, respectively:

$$u(t) \simeq u_n(t) = \sum_{i=0}^n a_i \psi_i(t) = \mathbf{A}^T \Psi(t), \quad (11)$$

$$u(t) \simeq u_n(t) = \sum_{i=0}^n \bar{a}_i \phi_i(t) = \bar{\mathbf{A}}^T \Phi(t), \quad (12)$$

so that

$$\Psi(t) = [\psi_0(t), \psi_1(t), \dots, \psi_n(t)]^T, \quad \Phi(t) = [\phi_0(t), \phi_1(t), \dots, \phi_n(t)]^T, \quad (13)$$

and

$$\mathbf{A} = [a_0, a_1, \dots, a_n]^T, \quad \bar{\mathbf{A}} = [\bar{a}_0, \bar{a}_1, \dots, \bar{a}_n]^T, \quad (14)$$

75 where

$$a_i = \bar{a}_i = u(ih), \quad i = 0, \dots, n. \quad (15)$$

3. Operational matrices of MHFs & QHFs

The purpose of this section is to present the fractional-order integral operational matrices based on the modified hat and quasi-hat functions.

3.1. Fractional order operational matrix of integration

80 Let us to state the following theorems:

Theorem 1. Suppose $\Psi(t)$ is given by (13) and $\alpha > 0$, then

$$I_t^\alpha \Psi(t) \simeq P^\alpha \Psi(t). \quad (16)$$

P^α is the $(n+1) \times (n+1)$ operational matrix of the Riemann-Liouville integral, illustrated as follows:

$$P^{(\alpha)} = \frac{h^\alpha}{2\Gamma(\alpha+3)} \begin{pmatrix} 0 & \zeta_1 & \zeta_2 & \zeta_3 & \zeta_4 & \zeta_5 & \zeta_6 & \dots & \zeta_{n-1} & \zeta_n \\ 0 & \beta_1 & \beta_2 & \beta_3 & \beta_4 & \beta_5 & \beta_6 & \dots & \beta_{n-1} & \beta_n \\ 0 & \eta_1 & \eta_2 & \eta_3 & \eta_4 & \eta_5 & \eta_6 & \dots & \eta_{n-1} & \eta_n \\ 0 & 0 & 0 & \beta_1 & \beta_2 & \beta_3 & \beta_4 & \dots & \beta_{n-3} & \beta_{n-2} \\ 0 & 0 & 0 & \eta_1 & \eta_2 & \eta_3 & \eta_4 & \dots & \eta_{n-3} & \eta_{n-2} \\ 0 & 0 & 0 & 0 & 0 & \beta_1 & \beta_2 & \dots & \beta_{n-5} & \beta_{n-4} \\ 0 & 0 & 0 & 0 & 0 & \eta_1 & \eta_2 & \dots & \eta_{n-5} & \eta_{n-4} \\ 0 & 0 & 0 & 0 & 0 & \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & \dots & \beta_1 & \beta_2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & \dots & \eta_1 & \eta_2 \end{pmatrix}, \quad (17)$$

where

$$\begin{aligned} \zeta_1 &= \alpha(2\alpha+3), \\ \zeta_k &= \left(k^{\alpha+1}(2k-3\alpha-6) + 2k^\alpha(\alpha+1)(\alpha+2) + (k-2)^{\alpha+1}(2-2k-\alpha) \right), \\ k &= 2, 3, \dots, n, \\ \beta_1 &= 4(\alpha+1) \\ \beta_k &= 4 \left((k-2)^{\alpha+1}(k+\alpha) + k^{\alpha+1}(2+\alpha-k) \right), \quad k = 2, 3, \dots, n, \\ \eta_k &= -\alpha\delta_{1k} + (2)^{\alpha+1}(2-\alpha)\delta_{2k} + ((3)^{\alpha+1}(4-\alpha) - 6(2+\alpha))\delta_{3k}, \quad k = 1, 2, 3, \\ \eta_k &= (k-4)^{\alpha+1}(6-2k-\alpha) - 6(k-2)^{\alpha+1}(2+\alpha) + (k)^{\alpha+1}(2k-2-\alpha), \\ k &= 4, \dots, n, \end{aligned} \quad (18)$$

85 where δ_{ij} is the Kronecker delta.

Proof. For the poof and more details see [16].

Theorem 2. Let $\Phi(t)$ be given by (13) and $\alpha > 0$, then

$$I_t^\alpha \Phi(t) \simeq Q^\alpha \Phi(t). \quad (19)$$

So that Q^α is the $(n+1) \times (n+1)$ operational matrix of order α , which is resulted by defining the Riemann-Liouville integral, and is generally represented
 90 as follows

$$Q^{(\alpha)} = \frac{h^\alpha}{2\Gamma(\alpha+3)} \begin{pmatrix} 0 & \rho_1 & \rho_2 & \rho_3 & \rho_4 & \cdots & \rho_{n-1} & \rho_n \\ 0 & \sigma_1 & \sigma_2 & \sigma_3 & \sigma_4 & \cdots & \sigma_{n-1} & \sigma_n \\ 0 & 0 & 0 & \rho_1 & \rho_2 & \cdots & \rho_{n-3} & \rho_{n-2} \\ 0 & 0 & 0 & \sigma_1 & \sigma_2 & \cdots & \sigma_{n-3} & \sigma_{n-2} \\ 0 & 0 & 0 & 0 & 0 & \cdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & \cdots & \rho_1 & \rho_2 \\ 0 & 0 & 0 & 0 & 0 & & \sigma_1 & \sigma_2 \\ 0 & 0 & 0 & 0 & 0 & \cdots & 0 & 0 \end{pmatrix}, \quad (20)$$

where

$$\begin{aligned} \rho_1 &= \alpha(2\alpha+3), \\ \rho_k &= \left(k^{\alpha+1}(2k-3\alpha-6) + 2k^\alpha(\alpha+1)(\alpha+2) + (k-2)^{\alpha+1}(2-2k-\alpha) \right), \\ k &= 2, 3, \dots, n, \\ \sigma_1 &= 3\alpha+4, \\ \sigma_k &= (k-2)^{\alpha+1}(2k+\alpha-2) - 2(k-2)^\alpha(2+\alpha)(1+\alpha) - (k)^{\alpha+1}(2k-6-3\alpha), \\ k &= 2, 3, \dots, n. \end{aligned} \quad (21)$$

Proof. For the proof and more details see [18].

If we approximate a function with MHFs and QHFs, we can use (16) and (19), to obtain the following result:

$$I_t^\alpha u(t) \simeq I_t^\alpha \left(\sum_{i=0}^n a_i \psi_i(t) \right) \simeq I_t^\alpha (A^T \Psi(t)) \simeq A^T P^\alpha \Psi(t). \quad (22)$$

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$$I_t^\alpha u(t) \simeq I_t^\alpha \left(\sum_{i=0}^n \bar{a}_i \phi_i(t) \right) \simeq I_t^\alpha (\bar{A}^T \Phi(t)) \simeq \bar{A}^T Q^\alpha \Phi(t). \quad (23)$$

4. Error analysis

In this section, quasi-hat and modified hat functions are proposed to estimate the absolute errors for approximating an arbitrary function.

According to reference [18], the absolute error when approximating an arbitrary function using QHFs (12), can be written as

$$|u(t) - u_n(t)| \simeq \frac{h}{2} \left| (j - k)(2 + (-1)^k + k - j)u'(kh) \right|, \quad (24)$$

wherein $j = t/h$, $t \in (kh, (k + 1)h)$, $k = 0, 1, 2, \dots, n$.

An analysis is performed to approximate an arbitrary function with MHFs (11), as follows [16]:

$$u_n(t) \simeq u(kh) + (t - kh)u'(kh) + \frac{(t - kh)^2}{2}u''(kh). \quad (25)$$

Also, the absolute error at the points $t \in (kh, (k + 1)h)$ is as follows

$$|u(t) - u_n(t)| \simeq \left| u(t) - u(kh) - (t - kh)u'(kh) - \frac{(t - kh)^2}{2}u''(kh) \right|. \quad (26)$$

For $t \in (kh, (k + 1)h)$, $k = 0, 1, 2, \dots, n$, $h \rightarrow 0$, obtain

$$|u(t) - u_n(t)| \simeq \frac{1}{2} \left| (jh - kh)^2 u''(kh) \right| = \frac{h^2}{2} \left| (j - k)^2 u''(kh) \right|. \quad (27)$$

where $j = t/h$. The absolute error of QHFs and MHFs for $j \rightarrow k$, $k = 0, \dots, n$, results

$$|u(kh) - u_n(kh)| \simeq 0, \quad (28)$$

thus $\forall k$, while $h \rightarrow 0$ or $n \rightarrow \infty$, result in

$$|u(t) - u_n(t)| \rightarrow 0. \quad (29)$$

5. Numerical process

In this section, two numerical approaches are presented for the solution of Eq.(1). As well recalling, the definition of Riemann-Liouville integral operator

of order α is as follows [15]

$$I_t^\alpha u(t) = \frac{1}{\Gamma(\alpha)} \int_0^t (t-\tau)^{\alpha-1} u(\tau) d\tau. \quad (30)$$

We apply Eq.(30) to both sides of the original equation Eq.(1). Then, using the properties of the Riemann-Liouville integral operator and the Caputo fractional
115 differential operator, and the initial conditions of Eq.(1), we obtain

$$\begin{aligned} u_1(t) &= \lambda_1 + I_t^\alpha p(t) + \frac{F_{13}}{V_3} I_t^\alpha u_3(t) - \frac{F_{31}}{V_1} I_t^\alpha u_1(t) - \frac{F_{21}}{V_1} I_t^\alpha u_1(t), \\ u_2(t) &= \lambda_2 + \frac{F_{21}}{V_1} I_t^\alpha u_1(t) - \frac{F_{32}}{V_2} I_t^\alpha u_2(t), \\ u_3(t) &= \lambda_3 + \frac{F_{31}}{V_1} I_t^\alpha u_1(t) + \frac{F_{32}}{V_2} I_t^\alpha u_2(t) - \frac{F_{13}}{V_3} I_t^\alpha u_3(t). \end{aligned} \quad (31)$$

5.1. Description of numerical algorithm based on MHFs

As described in Section 2, we can approximate all the functions in Eqs (31) with MHFs (11), as follows:

$$u_v(t) \simeq \sum_{i=0}^n a_{vi} \psi_i(t) = A_v^T \Psi(t), \quad v = 1, 2, 3, \quad (32)$$

$$\lambda_v \simeq \lambda_v \sum_{i=0}^n \psi_i(t) = \lambda_v E^T \Psi(t), \quad v = 1, 2, 3, \quad (33)$$

$$p(t) \simeq \sum_{i=0}^n p(ih) \psi_i(t) = G^T \Psi(t), \quad (34)$$

120 so that

$$A_v = [a_{v0}, a_{v1}, \dots, a_{vn}]^T, \quad v = 1, 2, 3, \quad (35)$$

$$E = [1, 1, \dots, 1]^T, \quad (36)$$

$$G = [p(0), p(h), \dots, p(nh)]^T. \quad (37)$$

Combining (16), (22) and substitution (32-34) into Eqs.(31) results in

$$\begin{aligned} A_1^T \Psi(t) &= \lambda_1 E^T \Psi(t) + G^T P^\alpha \Psi(t) + \frac{F_{13}}{V_3} A_3^T P^\alpha \Psi(t) - \frac{F_{31}}{V_1} A_1^T P^\alpha \Psi(t) \\ &\quad - \frac{F_{21}}{V_1} A_1^T P^\alpha \Psi(t), \\ A_2^T \Psi(t) &= \lambda_2 E^T \Psi(t) + \frac{F_{21}}{V_1} A_1^T P^\alpha \Psi(t) - \frac{F_{32}}{V_2} A_2^T P^\alpha \Psi(t), \end{aligned}$$

$$\begin{aligned}
A_3^T \Psi(t) &= \lambda_3 E^T \Psi(t) + \frac{F_{31}}{V_1} A_1^T P^\alpha \Psi(t) + \frac{F_{32}}{V_2} A_2^T P^\alpha \Psi(t) \\
&\quad - \frac{F_{13}}{V_3} A_3^T P^\alpha \Psi(t).
\end{aligned} \tag{38}$$

Also, we can write

$$\begin{aligned}
A_1^T - \lambda_1 E^T - G^T P^\alpha - \frac{F_{13}}{V_3} A_3^T P^\alpha + \frac{F_{31}}{V_1} A_1^T P^\alpha + \frac{F_{21}}{V_1} A_1^T P^\alpha &= 0, \\
A_2^T - \lambda_2 E^T - \frac{F_{21}}{V_1} A_1^T P^\alpha + \frac{F_{32}}{V_2} A_2^T P^\alpha &= 0, \\
A_3^T - \lambda_3 E^T - \frac{F_{31}}{V_1} A_1^T P^\alpha - \frac{F_{32}}{V_2} A_2^T P^\alpha + \frac{F_{13}}{V_3} A_3^T P^\alpha &= 0.
\end{aligned} \tag{39}$$

This system contains 3 (n+1) equations with 3 (n+1) unknown MHFs coefficients. Assume $P^\alpha = [\mu]_{ij}$, $i, j = 0, \dots, n$, the following results are obtained from the arrangement of elements in the operational matrix (17):

$$\begin{aligned}
[\mu_{ij}]_{i=0}^n &= 0, & j &= 0, \\
[\mu_{ij}]_{i=j+2}^n &= 0, & j &= 1, 3, \dots, n-1, \\
[\mu_{ij}]_{i=j+1}^n &= 0, & j &= 2, 4, \dots, n.
\end{aligned} \tag{40}$$

Based on (15), we get

$$a_{10} = \lambda_1, \quad a_{20} = \lambda_2, \quad a_{30} = \lambda_3. \tag{41}$$

As a result of the properties of the operational matrix of MHFs, Eqs (17) and (40), we introduce a recursive system($k/2$), $k = 2, 4, \dots, n$. To define system(1), take $P^\alpha = [\mu]_{ij}$, $i = 0, 1, k$, $j = 1, k$, and $k = 2$, so Eq.(39) is expressed as

follows:

$$\text{system}(1) : \begin{cases} a_{11} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^2 \mu_{i1} a_{3i} \right] - \left[\sum_{i=0}^2 p(ih) \mu_{i1} \right] + \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^2 \mu_{i1} a_{1i} \right] - \lambda_1 = 0, \\ a_{12} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^2 \mu_{i2} a_{3i} \right] - \left[\sum_{i=0}^2 p(ih) \mu_{i2} \right] + \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^2 \mu_{i2} a_{1i} \right] - \lambda_1 = 0, \\ a_{21} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^2 \mu_{i1} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^2 \mu_{i1} a_{2i} \right] - \lambda_2 = 0, \\ a_{22} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^2 \mu_{i2} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^2 \mu_{i2} a_{2i} \right] - \lambda_2 = 0, \\ a_{31} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^2 \mu_{i1} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^2 \mu_{i1} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^2 \mu_{i1} a_{3i} \right] - \lambda_3 = 0, \\ a_{32} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^2 \mu_{i2} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^2 \mu_{i2} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^2 \mu_{i2} a_{3i} \right] - \lambda_3 = 0. \end{cases}$$

Solving system(1) which includes 6 equations, coefficients of $\{a_{11}, a_{12}, a_{21}, a_{22}, a_{31}, a_{32}\}$ can be calculated. Then for $k = 4$, one has

$$\text{system}(2) : \begin{cases} a_{13} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^4 \mu_{i3} a_{3i} \right] - \left[\sum_{i=0}^4 p(ih) \mu_{i3} \right] + \left(\frac{F_{31}+F_{21}}{V_1} \right) \left[\sum_{i=0}^4 \mu_{i3} a_{1i} \right] - \lambda_1 = 0, \\ a_{14} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^4 \mu_{i4} a_{3i} \right] - \left[\sum_{i=0}^4 p(ih) \mu_{i4} \right] + \left(\frac{F_{31}+F_{21}}{V_1} \right) \left[\sum_{i=0}^4 \mu_{i4} a_{1i} \right] - \lambda_1 = 0, \\ a_{23} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^4 \mu_{i3} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^4 \mu_{i3} a_{2i} \right] - \lambda_2 = 0, \\ a_{24} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^4 \mu_{i4} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^4 \mu_{i4} a_{2i} \right] - \lambda_2 = 0, \\ a_{33} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^4 \mu_{i3} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^4 \mu_{i3} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^4 \mu_{i3} a_{3i} \right] - \lambda_3 = 0, \\ a_{34} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^4 \mu_{i4} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^4 \mu_{i4} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^4 \mu_{i4} a_{3i} \right] - \lambda_3 = 0, \end{cases}$$

unknown parameters $\{a_{13}, a_{14}, a_{23}, a_{24}, a_{33}, a_{34}\}$ are obtained by solving system(2).

Then we continue the successive process

135 \vdots

eventually, for $k = n$, result in

$$\text{system} \left(\frac{n}{2} \right) : \begin{cases} a_{1n-1} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^n \mu_{i n-1} a_{3i} \right] - \left[\sum_{i=0}^n p(ih) \mu_{i n-1} \right] + \left(\frac{F_{31}+F_{21}}{V_1} \right) \left[\sum_{i=0}^n \mu_{i n-1} a_{1i} \right] - \lambda_1 = 0, \\ a_{1n} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^n \mu_{i n} a_{3i} \right] - \left[\sum_{i=0}^n p(ih) \mu_{i n} \right] + \left(\frac{F_{31}+F_{21}}{V_1} \right) \left[\sum_{i=0}^n \mu_{i n} a_{1i} \right] - \lambda_1 = 0, \\ a_{2n-1} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^n \mu_{i n-1} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^n \mu_{i n-1} a_{2i} \right] - \lambda_2 = 0, \\ a_{2n} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^n \mu_{i n} a_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^n \mu_{i n} a_{2i} \right] - \lambda_2 = 0, \\ a_{3n-1} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^n \mu_{i n-1} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^n \mu_{i n-1} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^n \mu_{i n-1} a_{3i} \right] - \lambda_3 = 0, \\ a_{3n} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^n \mu_{i n} a_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^n \mu_{i n} a_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^n \mu_{i n} a_{3i} \right] - \lambda_3 = 0, \end{cases}$$

by solving system($\frac{n}{2}$) and finding the six unknown coefficients of this system, all coefficients are completely determined, and we can get the approximate solutions

140 $u_1(t), u_2(t), u_3(t)$, (32). The proposed approach reduces the original system to $\frac{n}{2}$ systems, which includes six algebraic equations.

5.2. Description of numerical algorithm based on QHFs

To obtain numerical solutions of Eqs. (31) using QHFs (12), we have

$$u_v(t) \simeq \sum_{i=0}^n \bar{a}_{vi} \phi_i(t) = \bar{A}_v^T \Phi(t), \quad v = 1, 2, 3, \quad (45)$$

$$\lambda_v \simeq \lambda_v \sum_{i=0}^n \phi_i(t) = \lambda_v E^T \Phi(t), \quad v = 1, 2, 3, \quad (46)$$

$$p(t) \simeq \sum_{i=0}^n p(ih) \phi_i(t) = G^T \Phi(t), \quad (47)$$

145 wherein

$$\bar{A}_v = [\bar{a}_{v0}, \bar{a}_{v1}, \dots, \bar{a}_{vn}]^T, \quad v = 1, 2, 3, \quad (48)$$

$$E = [1, 1, \dots, 1]^T, \quad G = [p(0), p(h), \dots, p(nh)]^T.$$

By applying (19-20) and (45-47) in Eqs. (31) the following results are obtained:

$$\begin{aligned} \bar{A}_1^T - \lambda_1 E^T - G^T Q^\alpha - \frac{F_{13}}{V_3} \bar{A}_3^T Q^\alpha + \frac{F_{31}}{V_1} \bar{A}_1^T Q^\alpha + \frac{F_{21}}{V_1} \bar{A}_1^T Q^\alpha &= 0, \\ \bar{A}_2^T - \lambda_2 E^T - \frac{F_{21}}{V_1} \bar{A}_1^T Q^\alpha + \frac{F_{32}}{V_2} \bar{A}_2^T Q^\alpha &= 0, \\ \bar{A}_3^T - \lambda_3 E^T - \frac{F_{31}}{V_1} \bar{A}_1^T Q^\alpha - \frac{F_{32}}{V_2} \bar{A}_2^T Q^\alpha + \frac{F_{13}}{V_3} \bar{A}_3^T Q^\alpha &= 0, \end{aligned} \quad (49)$$

the dimension of this system is $3(n+1) \times 3(n+1)$.

Suppose $Q^\alpha = [\theta]_{ij}$, $i, j = 0, \dots, n$. As shown in the operational matrix (20), we have

$$\begin{aligned} [\theta_{ij}]_{i=0}^n &= 0, & j &= 0, \\ [\theta_{ij}]_{i=0}^n &= 0, & j &= n, \\ [\theta_{ij}]_{i=j+1}^n &= 0, & j &= 1, 3, \dots, n-1, \\ [\theta_{ij}]_{i=j}^n &= 0, & j &= 2, 4, \dots, n. \end{aligned} \quad (50)$$

150 Using Eq. (15) and initial values, we get the following results:

$$\bar{a}_{10} = \lambda_1, \quad \bar{a}_{20} = \lambda_2, \quad \bar{a}_{30} = \lambda_3. \quad (51)$$

According to Eqs. (20) and (50), we introduce recursive system($k/2$), $k = 2, 4, \dots, n$. Thus, the following system is an appropriate representation to write (49) based on (50).

$$\text{system}(1) : \begin{cases} \bar{a}_{11} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{3i} \right] - \left[\sum_{i=0}^1 p(ih) \theta_{i1} \right] + \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{1i} \right] - \lambda_1 = 0, \\ \bar{a}_{21} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{2i} \right] - \lambda_2 = 0, \\ \bar{a}_{31} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^1 \theta_{i1} \bar{a}_{3i} \right] - \lambda_3 = 0, \end{cases}$$

by solving system(1) which includes 3 equations, coefficients of $\{\bar{a}_{11}, \bar{a}_{21}, \bar{a}_{31}\}$

155 can be calculated, then we get $\{\bar{a}_{12}, \bar{a}_{22}, \bar{a}_{32}\}$, as follows

$$\begin{aligned} \bar{a}_{12} &= \frac{F_{13}}{V_3} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{3i} \right] + \left[\sum_{i=0}^1 p(ih) \theta_{i2} \right] - \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{1i} \right] + \lambda_1, \\ \bar{a}_{22} &= \frac{F_{21}}{V_1} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{2i} \right] + \lambda_2, \\ \bar{a}_{32} &= \frac{F_{31}}{V_1} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{2i} \right] - \frac{F_{13}}{V_3} \left[\sum_{i=0}^1 \theta_{i2} \bar{a}_{3i} \right] + \lambda_3, \end{aligned}$$

then, for $k = 4$, one has

$$\text{system}(2) : \begin{cases} \bar{a}_{13} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{3i} \right] - \left[\sum_{i=0}^3 p(ih) \theta_{i3} \right] + \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{1i} \right] - \lambda_1 = 0, \\ \bar{a}_{23} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{2i} \right] - \lambda_2 = 0, \\ \bar{a}_{33} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^3 \theta_{i3} \bar{a}_{3i} \right] - \lambda_3 = 0, \end{cases}$$

unknown parameters $\{\bar{a}_{13}, \bar{a}_{23}, \bar{a}_{33}\}$ are calculated by solving system(2), then

we find $\{\bar{a}_{14}, \bar{a}_{24}, \bar{a}_{34}\}$, as follows

$$\begin{aligned} \bar{a}_{14} &= \frac{F_{13}}{V_3} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{3i} \right] + \left[\sum_{i=0}^3 p(ih) \theta_{i4} \right] - \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{1i} \right] + \lambda_1, \\ \bar{a}_{24} &= \frac{F_{21}}{V_1} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{2i} \right] + \lambda_2, \\ \bar{a}_{34} &= \frac{F_{31}}{V_1} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{2i} \right] - \frac{F_{13}}{V_3} \left[\sum_{i=0}^3 \theta_{i4} \bar{a}_{3i} \right] + \lambda_3, \end{aligned}$$

160 then we continue the process

⋮

finally, for $k = n$, result in

$$\text{system}\left(\frac{n}{2}\right) : \begin{cases} \bar{a}_{1n-1} - \frac{F_{13}}{V_3} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{3i} \right] - \left[\sum_{i=0}^{n-1} p(ih) \theta_{in-1} \right] + \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{1i} \right] - \lambda_1 = 0, \\ \bar{a}_{2n-1} - \frac{F_{21}}{V_1} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{2i} \right] - \lambda_2 = 0, \\ \bar{a}_{3n-1} - \frac{F_{31}}{V_1} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{2i} \right] + \frac{F_{13}}{V_3} \left[\sum_{i=0}^{n-1} \theta_{in-1} \bar{a}_{3i} \right] - \lambda_3 = 0, \end{cases}$$

having the coefficients $\{\bar{a}_{1n-1}, \bar{a}_{2n-1}, \bar{a}_{3n-1}\}$ from solving this system, the coefficients $\{\bar{a}_{1n}, \bar{a}_{2n}, \bar{a}_{3n}\}$ are found as follows

$$\begin{aligned} \bar{a}_{1n} &= \frac{F_{13}}{V_3} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{3i} \right] + \left[\sum_{i=0}^{n-1} p(ih) \theta_{in} \right] - \left(\frac{F_{31} + F_{21}}{V_1} \right) \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{1i} \right] + \lambda_1, \\ \bar{a}_{2n} &= \frac{F_{21}}{V_1} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{1i} \right] - \frac{F_{32}}{V_2} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{2i} \right] + \lambda_2, \\ \bar{a}_{3n} &= \frac{F_{31}}{V_1} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{1i} \right] + \frac{F_{32}}{V_2} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{2i} \right] - \frac{F_{13}}{V_3} \left[\sum_{i=0}^{n-1} \theta_{in} \bar{a}_{3i} \right] + \lambda_3. \end{aligned}$$

165 Once we determine the three unknown coefficients for system $\left(\frac{n}{2}\right)$, all the coefficients are determined, and we can get the approximate solutions $u_1(t)$, $u_2(t)$, and $u_3(t)$ (45). In this method, there is a need to solve $\frac{n}{2}$ systems, each of them consists of three linear algebraic equations, that can be done easily by direct methods. The computations are handled using the MATLAB package.

170 6. Convergence analysis of MHFs and QHFs approachers.

In this section, the convergence of the system (1) is examined based on the approximation of functions using MHFs and QHFs. Consider the system (31), which is equivalent to the system (1). The resulting equations must approximate the following equation.

$$\begin{aligned} E_{1n}^\alpha u(t) &= \left| u_1(t) - I_t^\alpha p(t) - \frac{F_{13}}{V_3} I_t^\alpha u_3(t) + \frac{F_{31}}{V_1} I_t^\alpha u_1(t) + \frac{F_{21}}{V_1} I_t^\alpha u_1(t) - \lambda_1 \right| \simeq 0, \\ E_{2n}^\alpha u(t) &= \left| u_2(t) - \frac{F_{21}}{V_1} I_t^\alpha u_1(t) + \frac{F_{32}}{V_2} I_t^\alpha u_2(t) - \lambda_2 \right| \simeq 0, \\ E_{3n}^\alpha u(t) &= \left| u_3(t) - \frac{F_{31}}{V_1} I_t^\alpha u_1(t) - \frac{F_{32}}{V_2} I_t^\alpha u_2(t) + \frac{F_{13}}{V_3} I_t^\alpha u_3(t) - \lambda_3 \right| \simeq 0. \quad (55) \end{aligned}$$

175 Here, two cases can be considered as follows:

(1) - Utilizing (27) to approximate the absolute error based on the functions

MHFs, results in

$$e(u_v(t)) \simeq \frac{h^2}{2} \left| (j-k)^2 u_v''(kh) \right|, \quad v = 1, 2, 3. \quad (56)$$

(2)- As a result of approximating the absolute error using (24), we obtain

$$\bar{e}(u_v(t)) \simeq \frac{h}{2} \left| (j-k)(2 + (-1)^k + k-j) u_v'(kh) \right|, \quad v = 1, 2, 3, \quad (57)$$

where $j = t/h$, and $t \in (kh, (k+1)h)$, $k = 0, 1, 2, \dots, n$.

180 For MHFs, utilizing (55) and (56), we get

$$\begin{aligned} E_{1n}^\alpha e(t) &\leq \frac{dh^2}{2} \left(|u_1''(t)| + b \left(|p''(t)| + \left| \frac{F_{13}}{V_3} u_3''(t) \right| + \left| \frac{F_{31}}{V_1} u_1''(t) \right| + \left| \frac{F_{21}}{V_1} u_1''(t) \right| \right) \right) \leq C_1 h^2, \\ E_{2n}^\alpha e(t) &\leq \frac{dh^2}{2} \left(|u_2''(t)| + b \left(\left| \frac{F_{21}}{V_1} u_1''(t) \right| + \left| \frac{F_{32}}{V_2} u_2''(t) \right| \right) \right) \leq C_2 h^2, \\ E_{3n}^\alpha e(t) &\leq \frac{dh^2}{2} \left(|u_3''(t)| + b \left(\left| \frac{F_{31}}{V_1} u_1''(t) \right| + \left| \frac{F_{32}}{V_2} u_2''(t) \right| + \left| \frac{F_{13}}{V_3} u_3''(t) \right| \right) \right) \leq C_3 h^2, \end{aligned}$$

wherein

$$d = \sup \left\{ (j-k)^2 \right\}, \quad b = \sup_{t, \tau \in [0, T]} \left| \frac{(t-\tau)^{\alpha-1}}{\Gamma(\alpha)} \right|, \quad \text{and as } h \rightarrow 0, \quad E_{vn}^\alpha e(t) \rightarrow 0, \\ v = 1, 2, 3.$$

For QHFs, when we take (55) and (57), we get

$$\begin{aligned} E_{1n}^\alpha \bar{e}(t) &\leq \frac{\bar{d}h}{2} \left(|u_1'(t)| + b \left(|p'(t)| + \left| \frac{F_{13}}{V_3} u_3'(t) \right| + \left| \frac{F_{31}}{V_1} u_1'(t) \right| + \left| \frac{F_{21}}{V_1} u_1'(t) \right| \right) \right) \leq \bar{C}_1 h, \\ E_{2n}^\alpha \bar{e}(t) &\leq \frac{\bar{d}h}{2} \left(|u_2'(t)| + b \left(\left| \frac{F_{21}}{V_1} u_1'(t) \right| + \left| \frac{F_{32}}{V_2} u_2'(t) \right| \right) \right) \leq \bar{C}_2 h, \\ E_{3n}^\alpha \bar{e}(t) &\leq \frac{\bar{d}h}{2} \left(|u_3'(t)| + b \left(\left| \frac{F_{31}}{V_1} u_1'(t) \right| + \left| \frac{F_{32}}{V_2} u_2'(t) \right| + \left| \frac{F_{13}}{V_3} u_3'(t) \right| \right) \right) \leq \bar{C}_3 h, \end{aligned}$$

185 wherein

$$\bar{d} = \sup \left\{ (j-k)(2 + (-1)^k + k-j) \right\}, \quad b = \sup_{t, \tau \in [0, T]} \left| \frac{(t-\tau)^{\alpha-1}}{\Gamma(\alpha)} \right|, \quad \text{and while } h \rightarrow 0, \\ E_{vn}^\alpha \bar{e}(t) \rightarrow 0, \quad v = 1, 2, 3.$$

7. Numerical experiments and discussions

In this section, three different pollution input models are considered to evaluate the effectiveness of the presented methods.

190

Example 1. Periodic Input Model

The model considers pollutants enter periodically into Lake 1. Let us consider $p(t) = a\sin(\omega t) + c$, where a and ω denote amplitude and frequency of the variation, respectively, and the average concentration of pollutants is represented
 195 by c . We assume Eq.(1) with, $a=\omega=c=1$,

$$\begin{aligned} {}_0^C D_t^\alpha u_1(t) &= 1 + \sin(t) + \frac{38}{1180} u_3(t) - \frac{20}{2900} u_1(t) - \frac{18}{2900} u_1(t), \\ {}_0^C D_t^\alpha u_2(t) &= \frac{18}{2900} u_1(t) - \frac{18}{850} u_2(t), & 0 < \alpha \leq 1, \quad t \in [0, T], \\ {}_0^C D_t^\alpha u_3(t) &= \frac{20}{2900} u_1(t) + \frac{18}{850} u_2(t) - \frac{38}{1180} u_3(t), \end{aligned} \quad (58)$$

with the initial conditions $u_1(0) = 0$, $u_2(0) = 0$, and $u_3(0) = 0$.

There have been several studies of this problem (58). For the pollution monitoring, results of the presented methods and some other methods for one year are shown in Tables 2-5. Similar numerical results are obtained from the
 200 two proposed methods in Table 2 clarifying that water pollution in all three lakes rises as time progresses. Figure 3(a) demonstrates the sinusoidal behavior of lake 1. In this case, in analyzing Tables and plots, we find that lake 2 has low pollution levels compared to lake 3, and both lakes' pollution levels are increasing exponentially.

205 **Example 2. Exponential input model**

In this example, we will consider a system of fractional order differential equations (1) with $p(t) = ae^{bt}$ with $a = 1$, $b = 1$, $u_1(0) = 0$, $u_2(0) = 0$, and $u_3(0) = 0$, and other parameters the same as in example 1.

A numerical analysis is done for this case using MHFs and QHFs with $\alpha =$
 210 $0.5, 1$, (see Tables 6, 7). It seems that as alpha decreases from 1 down to 0.5, contamination increases on the time interval $[0, 1]$. The results show that the pollution in all three lakes exhibit exponential behavior. As expected, the pollution levels in all three lakes are increasing, although lake 2 appears to be lower pollution.

215 **Example 3. Impulse Input Model**

In this case, the contaminant has been released directly into lake 1, (input impact method). In this example, we will consider a system of fractional order differential equations (1) with $p(t) = 100$ and other parameters the same as in example 1.

220 Figure 4 shows the contamination results for different values of α based on MHFs and QHFs. According to Figure 4, the pollution behaviors of the three lakes using the MHFs and QHFs approaches are similar. When $\alpha = 1$, lake 1 behavior would reflect the effects of a constant concentration of pollution increasing over time. Also, for $\alpha = 1$, lakes 2 and 3 exhibit exponential pollution
225 growth.

8. Conclusion

In this paper, two algorithms are implemented to solve a fractional model of pollution in a system of lakes. Applying the modified hat functions and quasi-hat functions, algorithms have been developed for monitoring lakes' water pollution
230 for the first time. Analyzing the method's absolute errors and convergence are addressed. A comparison of the numerical results of two proposed algorithms with the results of other numerical approaches confirms both their efficiency and accuracy. This fractional system is solved using MHFs and QHFs with $n/2$ algebraic systems with dimensions of 6×6 and 3×3 , respectively. This is one
235 of the benefits of these two algorithms, so the methods proposed are simple and computationally efficient for large values of n . The amount of pollution in lake 1 is much higher than in lakes 2 and 3 because it is the principal source of pollutants. Generally, the contamination levels are inversely related to alpha at the beginning, and over time, the pollution levels and alpha values become di-
240 rectly related. A variety of similar problems can be addressed with the proposed methods, especially QHF, and we intend to investigate these issues.

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245 References

- [1] Biazar, J., Farrokhi, L. and Islam, M. “Modeling the pollution of a system of lakes”, *Appl. Math. Comput.*, **178**(2), pp. 423-430 (2006). doi:<https://doi.org/10.1016/j.amc.2005.11.056>.
- [2] Nadeem, S., Abbas Haider, J. and Akhtar, S. “Mathematical modeling of williamson’s model for blood flow inside permeable multiple stenosed
250 arteries with electro-osmosis”, *Sci. Iran.*, (2023). doi:<https://doi.org/10.24200/sci.2023.59837.6457>.
- [3] Biazar, J. and Ebrahimi, H. “Orthonormal bernstein polynomials for volterra integral equations of the second kind”, *Int. J. Appl. Math. Res.*,
255 **9**(1), pp. 9-20 (2019).
URL <https://www.sciencepubco.com/index.php/ijamr/article/view/29636>
- [4] Kumari, A. and Kukreja, V. K. “Study of 4th order kuramoto-sivashinsky equation by septic hermite collocation method”, *Appl. Numer. Math.*, **188**,
260 pp. 88–105 (2023). doi:<https://doi.org/10.1016/j.apnum.2023.03.001>.
- [5] Khader, M. “Numerical treatment for a nine-dimensional chaotic lorenz model with the rabotnov fractional-exponential kernel fractional derivative”, *Sci. Iran.* (2023).
265 URL http://scientiairanica.sharif.edu/article_23185_6318ef54f5b4f441c455a3c33e688869.pdf

- [6] Biazar, J., Shahbala, M. and Ebrahimi, H. “Vim for solving the pollution problem of a system of lakes”, *J. Control Sci. Eng.* 2010 (2010). doi:<https://doi.org/10.1155/2010/829152>.
- 270 [7] Yüzbaşı, Ş., Şahin, N. and Sezer, M. “A collocation approach to solving the model of pollution for a system of lakes”, *Math. Comput. Model.*, **55**(3), pp. 330-341 (2012). doi:<https://doi.org/10.1016/j.mcm.2011.08.007>.
- [8] Haq, E. U. “Analytical solution of fractional model of pollution for a system lakes”, *CRPASE: Transactions of Applied Sciences*, **06**(04), PP. 302-308, (2020).
 275 URL <http://www.crpase.com/archive/CRPASE-Vol-06-issue-04-89821872.pdf>
- [9] Khader, M., El Danaf, T. S. and Hendy, A. “A computational matrix method for solving systems of high order fractional differential equations”, *Appl. Math. Model.*, **37**(6), pp. 4035-4050 (2013).
 280 doi:<https://doi.org/10.1016/j.apm.2012.08.009>.
 URL <https://www.sciencedirect.com/science/article/pii/S0307904X12004659>
- [10] Prakasha, D. and Veerasha, P. “Analysis of lakes pollution model with mittag-leffler kernel”, *J. Ocean Eng. Sci.*, **5**(4), pp. 310-322 (2020). doi:
 285 <https://doi.org/10.1016/j.joes.2020.01.004>.
- [11] Hatipoğlu, V. F. “A novel model for the contamination of a system of three artificial lakes”, *Discrete Contin. Dyn. Syst.*, **14**(7), pp. 2261-2272 (2021). doi:<https://doi.org/10.3934/dcdss.2020176>.
- 290 [12] Ghosh, I., Chowdhury, M., Aznam, S. M. et al. “Measuring the pollutants in a system of three interconnecting lakes by the semianalytical method”, *J. Appl. Math.* 2021 (2021). doi:<https://doi.org/10.1155/2021/6664307>.
- [13] Shiri, B. and Baleanu, D. “A general fractional pollution model for lakes”,

- 295 *Commun. Appl. Math. Comput.*, **4**(3), pp. 1105-1130, (2022). doi:<https://doi.org/10.1007/s42967-021-00135-4>.
- [14] Yönet, N., Gürbüç, B. and Gökçe, A. “An alternative numerical approach for an improved ecological model of interconnected lakes with a fixed pollutant”, *Comput. Appl. Math.* **42**(1), 56 (2023) . doi:<https://doi.org/10.1007/s40314-023-02191-3>.
- 300 [15] Podlubny, I. “Fractionl differential equations”, mathematics in *science and engineering*, Academic Press, New York (1999).
URL https://books.google.co.in/books/about/Fractional_Differential_Equations.html?id=K5FdXohLto0C
- [16] Biazar, J. and Ebrahimi, H. “A numerical algorithm for a class of nonlinear fractional Volterra integral equations via modified hat functions”, *J. Integral Equ. Appl.*, **34**(3), pp. 295-316 (2022). doi:[10.1216/jie.2022.34.295](https://doi.org/10.1216/jie.2022.34.295).
305 URL <https://doi.org/10.1216/jie.2022.34.295>
- [17] Ebrahimi, H. and Biazar, J. “Cubic hat-functions approximation for linear and non-linear fractional integral-differential equations with weakly singular kernels”, *Iran. J. Numer. Anal. Optim.*, **13**(3), pp. 500-531 (2023).
310 doi:[10.22067/ijnao.2023.80249.1209](https://doi.org/10.22067/ijnao.2023.80249.1209).
- [18] Biazar, J. and Ebrahimi, H. “A one-step algorithm for strongly non-linear full fractional duffing equations”, *Comput. methods differ. equ.*, (2023).
315 doi:[10.22034/cmde.2023.53596.2256](https://doi.org/10.22034/cmde.2023.53596.2256).

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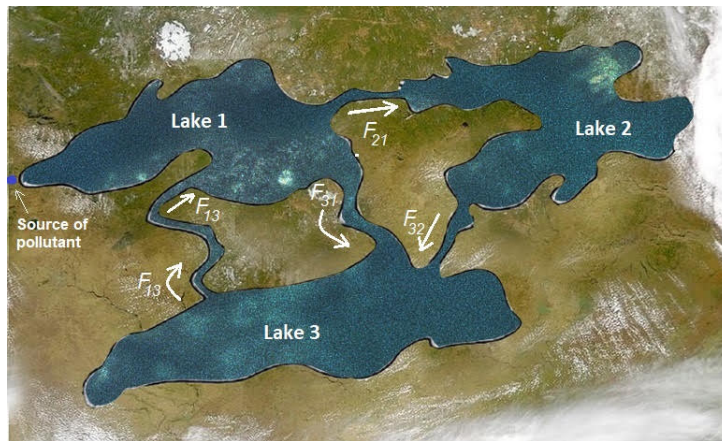


Figure 1: Aerial map of the arrangement of lakes with interconnecting channels.

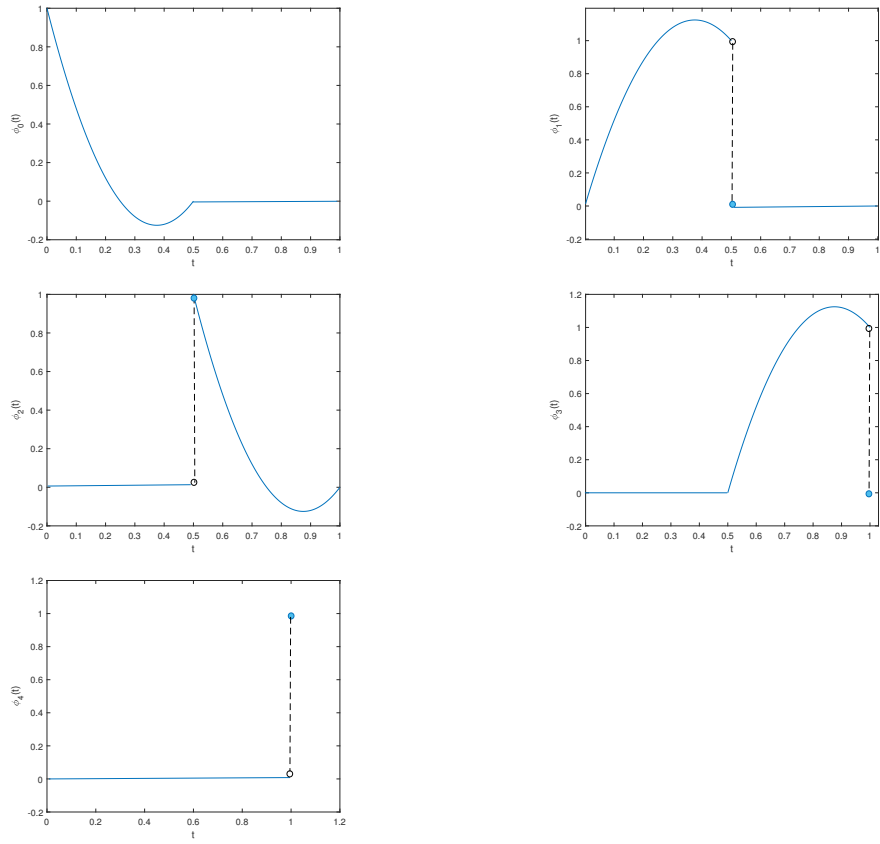
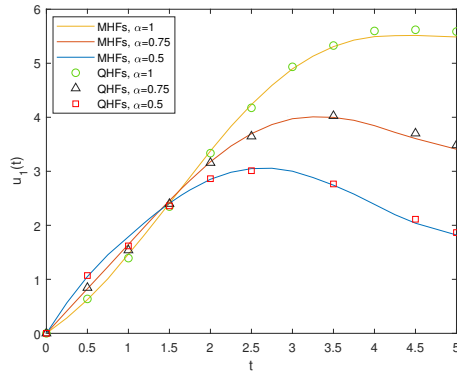
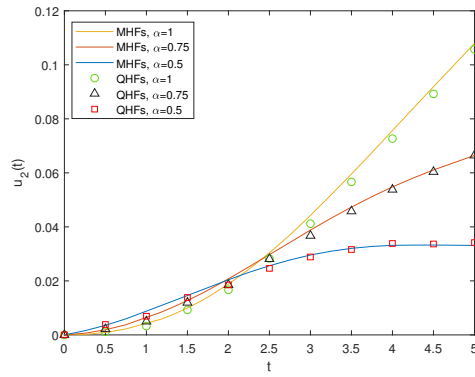


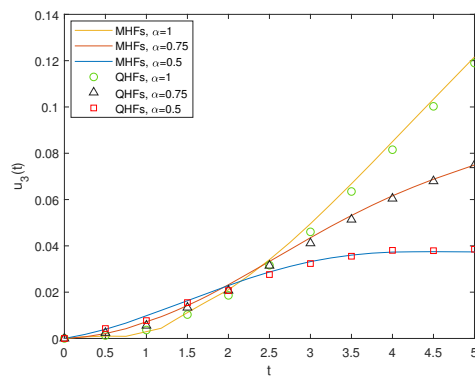
Figure 2: Plots of the QHFs, up to $n = 4, T = 1$.



(a) Lake 1

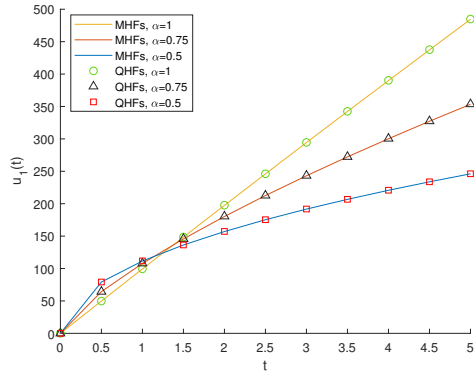


(b) Lake 2

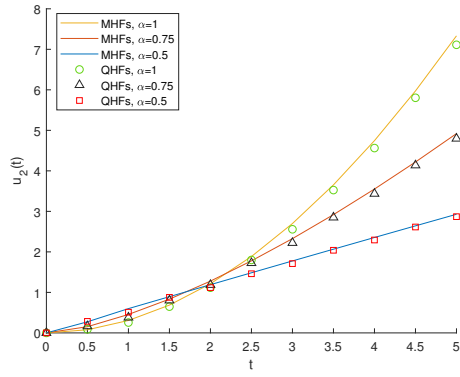


(c) Lake 3

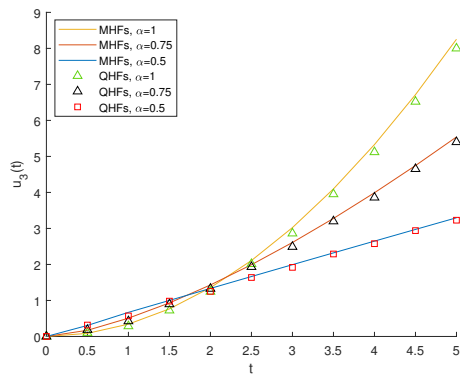
Figure 3: Distribution of pollution in the three Lakes by MHFs and QHFs, for example 1.



(a) Lake 1



(b) Lake 2



(c) Lake 3

Figure 4: The behavior of Pollution in each lake using MHFs and QHFs, for example 3.

Table 1: Nomenclature section of the paper.

Parameters and variables	Explanation	Unit
$u_1(t)$	The amount of the pollutant in lake 1 at any time t	ppm
$u_2(t)$	The amount of the pollutant in lake 2 at any time t	ppm
$u_3(t)$	The amount of the pollutant in lake 3 at any time t	ppm
$p(t)$	The rate at which the pollutant enters the first lake per unit of time t	ppm
V_1	The volume of water in the lake 1	m^3
V_2	The volume of water in the lake 2	m^3
V_3	The volume of water in the lake 3	m^3
F_{13}	The flow rate from lake 3 into lake 1	$m^3/year$
F_{21}	The flow rate from lake 1 into lake 2	$m^3/year$
F_{31}	The flow rate from lake 1 into lake 3	$m^3/year$
F_{32}	The flow rate from lake 3 into lake 2	$m^3/year$
t	time	year
T	period of time	year
λ_1	The initial amount of the pollutants in the lake 1	ppm
λ_2	The initial amount of the pollutants in the lake 2	ppm
λ_3	The initial amount of the pollutants in the lake 3	ppm

Table 2: Numerical results of Example 1 for three lakes, $\alpha = 1$, $n = 32$.

Time (in year)	MHFs			QHF's		
	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$
0.0	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
0.125	0.13269583	0.5044×10^{-4}	0.5606×10^{-4}	0.13205645	0.4602×10^{-4}	0.5115×10^{-4}
0.250	0.28064517	0.2094×10^{-3}	0.2329×10^{-3}	0.27937983	0.1996×10^{-3}	0.2220×10^{-3}
0.375	0.44346046	0.4886×10^{-3}	0.5433×10^{-3}	0.44159249	0.4724×10^{-3}	0.5254×10^{-3}
0.500	0.62051910	0.8986×10^{-3}	0.9997×10^{-3}	0.61808134	0.8755×10^{-3}	0.9738×10^{-3}
0.625	0.81097317	0.1450×10^{-2}	0.1613×10^{-2}	0.80800747	0.1418×10^{-2}	0.1578×10^{-2}
0.750	1.01376264	0.2152×10^{-2}	0.2396×10^{-2}	1.01031920	0.2112×10^{-2}	0.2351×10^{-2}
0.875	1.22763196	0.3014×10^{-2}	0.3356×10^{-2}	1.22374664	0.2975×10^{-2}	0.3312×10^{-2}
1.0	1.45114965	0.4044×10^{-2}	0.4504×10^{-2}	1.44690889	0.3994×10^{-2}	0.4449×10^{-2}

Table 3: Numerical results of the proposed and some other methods for the sinusoidal input of Lake 1 with $\alpha = 1$.

Time (in year)	MHF's	QHF's	Bessel polynomials	VPM
t	n=10	n=10	See[7], n=10	See[8]
0.2	0.219654473	0.216408198	0.219654467	-
0.4	0.477756712	0.471445300	0.477756680	-
0.5	0.620522784	0.614912667	-	0.62051
0.6	0.771858751	0.762787012	0.771858670	-
0.8	1.098057381	1.086641630	1.098057233	-
1.0	1.451149882	1.437901320	1.451149651	1.45115

Table 4: Numerical results of the proposed and some other methods for the sinusoidal input of Lake 2 with $\alpha = 1$.

Time (in year)	MHF's	QHF's	Bessel polynomials	VPM
t	n=10	n=10	See[7], n=10	See[8]
0.2	0.1321026×10^{-3}	0.1092542×10^{-3}	0.1320999×10^{-3}	-
0.4	0.5597490×10^{-3}	0.5062570×10^{-3}	0.5597436×10^{-3}	-
0.5	0.8986229×10^{-3}	0.8493274×10^{-3}	-	0.898×10^{-3}
0.6	0.1327956×10^{-2}	0.1236559×10^{-2}	0.1327949×10^{-2}	-
0.8	0.2477604×10^{-2}	0.2341872×10^{-2}	0.2477594×10^{-2}	-
1.0	0.4043740×10^{-2}	0.3858345×10^{-2}	0.4043728×10^{-2}	0.4043×10^{-2}

Table 5: Numerical results of the proposed and some other methods for the sinusoidal input of Lake 3 with $\alpha = 1$.

Time (in year)	MHF's	QHF's	Bessel polynomials	VPM
t	n=10	n=10	See[7], n=10	See[8]
0.2	0.1468580×10^{-3}	0.1214503×10^{-3}	0.1468549×10^{-3}	-
0.4	0.6225887×10^{-3}	0.5630246×10^{-3}	0.6225828×10^{-3}	-
0.5	0.9997562×10^{-3}	0.9448416×10^{-3}	-	0.999×10^{-3}
0.6	0.1477775×10^{-2}	0.1375878×10^{-2}	0.1477766×10^{-2}	-
0.8	0.2758474×10^{-2}	0.2606978×10^{-2}	0.2758463×10^{-2}	-
1.0	0.4504327×10^{-2}	0.4297170×10^{-2}	0.4504314×10^{-2}	0.4504×10^{-2}

Table 6: Numerical results of exponential input for three lakes, $\alpha = 0.5$, $n = 10$.

Time (in year)	MHFs			QHF's		
	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$
0.0	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
0.1	0.38034144	0.5862×10^{-3}	0.6525×10^{-3}	0.38303751	0.6187×10^{-3}	0.6889×10^{-3}
0.2	0.57477572	0.1346×10^{-2}	0.1500×10^{-2}	0.55180692	0.1114×10^{-2}	0.1240×10^{-2}
0.3	0.75333265	0.2125×10^{-2}	0.2369×10^{-2}	0.75026575	0.2070×10^{-2}	0.2307×10^{-2}
0.4	0.93190143	0.2991×10^{-2}	0.3336×10^{-2}	0.89904555	0.2712×10^{-2}	0.3023×10^{-2}
0.5	1.11720443	0.3944×10^{-2}	0.4400×10^{-2}	1.10960000	0.3834×10^{-2}	0.4277×10^{-2}
0.6	1.31353450	0.4997×10^{-2}	0.5576×10^{-2}	1.26999927	0.4642×10^{-2}	0.5179×10^{-2}
0.7	1.52400197	0.6157×10^{-2}	0.6873×10^{-2}	1.51170788	0.5989×10^{-2}	0.6987×10^{-2}
0.8	1.75156787	0.7438×10^{-2}	0.8305×10^{-2}	1.69561933	0.6987×10^{-2}	0.7800×10^{-2}
0.9	1.99882170	0.8852×10^{-2}	0.9886×10^{-2}	1.98126556	0.8616×10^{-2}	0.9619×10^{-2}
1.0	2.26867203	0.1041×10^{-1}	0.1163×10^{-1}	2.19794126	0.9843×10^{-2}	0.1099×10^{-1}

Table 7: Numerical results of exponential input for three lakes, $\alpha = 1$, $n = 10$.

Time (in year)	MHFs			QHF's		
	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$	Lake 1: $u_1(t)$	Lake 2: $u_2(t)$	Lake 3: $u_3(t)$
0.0	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000	0.00000000
0.1	0.10509874	0.3203×10^{-4}	0.3560×10^{-4}	0.10605399	0.3835×10^{-4}	0.4263×10^{-4}
0.2	0.22112299	0.1325×10^{-3}	0.1473×10^{-3}	0.21729710	0.1096×10^{-3}	0.1218×10^{-3}
0.3	0.34654789	0.3084×10^{-3}	0.3429×10^{-3}	0.34920205	0.2907×10^{-3}	0.3234×10^{-3}
0.4	0.49062641	0.5674×10^{-3}	0.5693×10^{-3}	0.48213712	0.5119×10^{-3}	0.5738×10^{-3}
0.5	0.64677541	0.9180×10^{-3}	0.1021×10^{-2}	0.63972225	0.8667×10^{-3}	0.9642×10^{-2}
0.6	0.81922527	0.1021×10^{-2}	0.1524×10^{-2}	0.80504959	0.1270×10^{-2}	0.1413×10^{-2}
0.7	1.00966072	0.1933×10^{-2}	0.2151×10^{-2}	0.99724405	0.1835×10^{-2}	0.2043×10^{-2}
0.8	1.22000692	0.2619×10^{-2}	0.2915×10^{-2}	1.19889530	0.2461×10^{-2}	0.2739×10^{-2}
0.9	1.45232162	0.3441×10^{-2}	0.3831×10^{-2}	1.43336338	0.3284×10^{-2}	0.3656×10^{-2}
1.0	1.70895624	0.4412×10^{-2}	0.4914×10^{-2}	1.67938243	0.4178×10^{-2}	0.4653×10^{-2}