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# A comparison between pre-Newton and post-Newton approaches for solving a physical singular second-order boundary problem in the semi-infinite interval

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In this paper, two numerical approaches based on the Newton iteration method with spectral algorithms are introduced to solve the Thomas-Fermi equation. That Thomas-Fermi equation is a nonlinear singular ordinary differential equation (ODE) with a boundary condition in infinite. In these schemes, the Newton method is combined with a spectral method where in one of those, by the Newton method we convert nonlinear ODE to a sequence of linear ODE and then, solve them using the spectral method. In another one, by the spectral method, the nonlinear ODE is converted to a system of nonlinear algebraic equations, then, this system is solved by the Newton method. In both approaches, the spectral method is based on the fractional order of rational Gegenbauer functions. Finally, the obtained results of the two introduced schemes are compared to each other in accuracy, runtime, and iteration number. Numerical experiments are presented showing that our methods are as accurate as the best results obtained until now. Copyright © 2022 Shahid Beheshti University.

**Keywords:** Pre-Newton method; Post-Newton method; Fractional order of rational Gegenbauer functions; Thomas-Fermi equation; Spectral method.

## 1. Introduction

Nonlinear problems arise in various fields of research such as biology, cognitive sciences, engineering, finance, etc. One of the branches of nonlinear problems is nonlinear ODEs which have unbounded domains. Since these problems are significant, many researchers developed different numerical schemes to solve them. There are various numerical algorithms to compute the solution of nonlinear problems over the semi-infinite domains such as the Adomian decomposition method [1–4], finite difference and finite element methods [5–7], Hermite collocation [8], meshless methods [9–11], etc. In this work, two different approaches based on the combination of spectral methods and Newton family algorithms are introduced. As the first one, we can refer to the pre-Newton. In the pre-Newton approach, a linearization method is done directly on the nonlinear ODE; then, nonlinear ODE is converted to a sequence of linear ODE which can be solved by different numerical algorithms such as spectral methods. The second approach is the post-Newton. In the post-Newton approach, by using a numerical method the nonlinear ODE is converted to a system of nonlinear algebraic equations, and then, this system is solved using various Newton-type algorithms. To show the efficiency of these approaches and compare them to each other, we consider a nonlinear ODE called the Thomas-Fermi equation which arises in theoretical physics as a test problem. This model has two significant roles in mathematical physics for two reasons: Thomas-Fermi equation was enhanced to model the effective nuclear charge in heavy atoms, and was investigated to analyze the potentials and charge densities of atoms having numerous electrons [12].

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**Table 1.** Different forms of Thomas-Fermi equation

Equation	Boundary conditions	Unknown	coordinate
$\frac{d^2y}{dx^2} - \frac{1}{\sqrt{x}}y^{\frac{3}{2}}(x) = 0$	$y(0) = 1, y(\infty) = 0$	$y$	$x \in [0, \infty]$
$z\{y\frac{d^2y}{dz^2} + \frac{dy}{dz}\} - y\frac{dy}{dz} - 2z^2y^3 = 0$	$y(0) = 1, y(\infty) = 0$	$\sqrt{y}$	$z = \sqrt{x}$
$z\frac{d^2y}{dz^2} - \frac{dy}{dz} - 4z^2y^{\frac{3}{2}} = 0$	$y(0) = 1, y(\infty) = 0$	$y$	$z = \sqrt{x}$

In this paper, pre-Newton and post-Newton approaches based on the fractional order of rational Gegenbauer (FRG) functions are used to solve the Thomas-Fermi equation in the semi-infinite interval. The main aim of this paper is to present a kind of collocation method based on FRG for the solving Thomas-Fermi equation which can obtain the most accurate results which are reported until now. In this paper, we are going to compute  $y'(0)$ . The obtained value for  $y'(0)$  is as follows:

$$-1.588071022611375312718684509423950109452746621674825616765677.$$

This value is obtained in [37] using 600 basis functions. But in this paper, we will obtain this value by using only 200 basis functions. The main advantage of the presented method is highly convergence rate of it. On the other hand, it has a good time efficiency. The main idea behind this paper is to use the linearization method to overcome the non-linearity of the Thomas-Fermi equation and we have used fractional order of rational Gegenbauer functions for the first time for solving the Thomas-Fermi equation. Moreover, the main advantage of our algorithm is the reduction of the essential basis function which is needed for obtaining the best possible accuracy. The organization of the paper is expressed as follows: Thomas-Fermi equation is introduced in Section 2, and the Gegenbauer polynomials and FRG functions are introduced in Section 3. Section 4 contains the Newton-Kantorovich method and the application of spectral methods. Results and discussion of the proposed methods are shown in Section 4. Finally, a conclusion is provided in Section 5.

## 2. Thomas-Fermi equation

The Thomas-Fermi theorem illustrates how the energy of an electronic system,  $E$ , and the electronic density,  $\rho$ , are connected to each other by the following formula [14]:

$$E[\rho] = \frac{9}{10B} \int \rho(r) d\tau + \frac{1}{2} \int \frac{\rho(r)\rho(r')}{|r-r'|} d\tau' d\tau + \int \rho(r)\nu(r) d\tau, \tag{1}$$

where  $\nu(r)$  is the external potential and  $B = 3(3\pi)^{-\frac{2}{3}}$ . In order to obtain the density the energy functional should be minimized with respect to  $\rho$  and subject to the normalization restriction  $\int \rho(r) d\tau = N$  where  $N$  is the number of electrons

$$\frac{3}{2B}\rho(r)^{\frac{3}{2}} + \int \frac{\rho(r')}{|r-r'|} d\tau' + \nu(r) = \mu, \tag{2}$$

where  $\mu$  is the Lagrange multiplier related to the normalization restriction [13, 14]. By using Poisson's equation to remove the density and a change of variables Thomas-Fermi equation is obtained as below:

$$\frac{d^2y}{dx^2} = \frac{1}{\sqrt{x}}y^{\frac{3}{2}}(x), \tag{3}$$

with the following boundary conditions:

$$y(0) = 1, \quad \lim_{x \rightarrow \infty} y(x) = 0. \tag{4}$$

This equation describes the charge density in atoms of high atomic number and appears in the problem of determining the effect of nuclear charge in heavy atoms [14–16].

As the solution of the Thomas-Fermi equation is effective in theoretical physics, many scientists have studied this model. Moreover, this equation has three different forms which can be effective on the rate of convergence of the using numerical algorithm [37]. These three forms of the Thomas-Fermi equation are listed in Table 1.

One special parameter in the Thomas-Fermi equation is the first derivative of the unknown function at the region  $y'(0)$ . This importance is because of some reasons, as the first one, we can refer to the expansion of  $y$  about the region, the expansion of  $y$  about the region is as follows [17]:

$$y(x) = 1 + \lambda x + \frac{4}{3}x^{\frac{3}{2}} + \frac{2\lambda}{5}x^{\frac{5}{2}} + \frac{1}{3}x^3 + \frac{3\lambda^2}{70}x^{\frac{7}{2}} + \dots, \tag{5}$$

**Table 2.** A brief bibliography on Thomas-Fermi equation

Years	Description
1930–1970	In these years, scientists studied the singularity and convergence of the Thomas-Fermi equation, found an analytical solution [17], and investigate the asymptotic behavior of $y(x)$ [19].
1970–2000	Researchers found an alternate analytical solution for the Thomas-Fermi equation using perturbative procedure [18], solved the Thomas-Fermi equation by standard decomposition method [1], Adomian decomposition method and Padé approximation [2,4].
2000-2010	In this decade, scientists proposed various approaches for approximating the solution of the Thomas-Fermi equation such as a combination of a semi-inverse scheme and the Ritz method [20], piecewise quasilinearization technique [21], an iterative approach and the sweep method [22], computing the potential slope at the origin by exploiting integral properties of the Thomas-Fermi equation [23], rational Chebyshev collocation method [24].
2010–2015	Scientists used semi-analytical and numerical approaches to solve the Thomas-Fermi equation with high accuracy. These techniques are improved Adomian decomposition method [25], optimal parametric iteration method [26], combination of three schemes based on Taylor series, Padé approximates and conformal mappings [27], the Hankel-Padé method [28], an adaptive finite element method based on moving mesh [12], Homotopy analysis method and Padé approximates [29], Newton-Kantorovich iteration and collocation approach based on rational Chebyshev functions [30], Sinc-collocation method [14], Rational second-kind Chebyshev pseudospectral technique [31], collocation method on Hermite polynomials [8]
2015–2018	Recently, researchers proposed fractional order of rational orthogonal [32,33] and non-orthogonal functions [34, 35] for approximating Thomas-Fermi equation. In 2018, Sabir et al suggest an artificial neural network [36] to solve that. Moreover, some other researchers study coordinate transformations [37] for approximating the Thomas-Fermi equation and found a highly accurate solution to 60 decimal places for $y'(0)$

where  $\lambda = y'(0) < 0$ . On the other hand,  $y'(0)$  can be used to obtain the energy of a neutral atom by the following formula:

$$E = \frac{6}{7} \left(\frac{4\pi}{3}\right)^{\frac{2}{3}} Z^{\frac{7}{3}} y'(0), \tag{6}$$

where  $Z$  is the nuclear charge [18].

As mentioned above Thomas-Fermi equation has special significance in theoretical physics and thanks to this importance many researchers develop various numerical algorithms to approximate the solution of the Thomas-Fermi equation. We summarize some previous works in the literature in Table 2.

### 3. Fractional order of rational Gegenbauer (FRG) functions

There are various types of orthogonal polynomials, which have different behaviors and properties. Choosing a good orthogonal function as a basis that behaves as same as the behavior of the exact solution is a challenging problem in spectral methods because we have no exact solution. But in some problems such as the Thomas-Fermi equation, although we have no exact solution we have some information about the behavior of the solution.

As mentioned in Eq. (5),  $y(x)$  can be expanded by a power series of  $x^{\frac{1}{2}}$  [33]. So if we choose fractional functions as a basis it can be fitted to the Baker expansion. Moreover, the Thomas-Fermi equation is defined in the semi-infinite domain. One choice for the semi-infinite domains is rational functions. Therefore, we select the fractional order of rational Gegenbauer functions as a basis. In this section, we introduce Gegenbauer polynomials and FRG functions, then we explain how to use the FRG function for function approximating.

#### 3.1. Gegenbauer polynomials

In this paper, we use the fractional order of the rational Gegenbauer function, where this function is obtained from Gegenbauer polynomials. The Gegenbauer polynomial of degree  $n$ ,  $G_n^a(x)$ , and order  $a > -\frac{1}{2}$  is solution of following differential equation:

$$(1 - x^2) \frac{d^2y}{dx^2} - (2a + 1)x \frac{dy}{dx} + n(n + 2a)y = 0, \tag{7}$$

where  $n$  is a positive integer.

The standard Gegenbauer polynomial  $G_n^a(x)$ , is defined as follows:

$$G_n^a(x) = \sum_{j=0}^{\lfloor \frac{n}{2} \rfloor} (-1)^j \frac{\Gamma(n+a-j)}{j!(n-2j)!\Gamma(a)} (2x)^{n-2j}, \tag{8}$$

where  $\Gamma(\cdot)$  is the Gamma function.

The Gegenbauer polynomials are orthogonal over the interval  $[-1, 1]$  with the weight function  $w(x) = (1-x^2)^{a-\frac{1}{2}}$  which means:

$$\int_{-1}^1 G_n^a(x)G_m^a(x)w(x)dx = \frac{\pi 2^{1-2a}\Gamma(n+2a)}{n!(n+a)(\Gamma(a))^2} \delta_{nm}, \tag{9}$$

where  $\delta_{nm}$  is the Kronecker delta function.

In addition, Gegenbauer polynomials can be obtained by the following recursive formula:

$$G_0^a(x) = 1, \quad G_1^a(x) = 2ax, \tag{10}$$

$$G_{n+1}^a(x) = \frac{1}{n+1} [2x(n+a)G_n^a(x) - (n+2a-1)G_{n-1}^a(x)], \quad n \geq 1 \tag{11}$$

### 3.2. Fractional order of rational Gegenbauer (FRG) functions

Scientists have been proposing the fractional order of rational functions such as rational Chebyshev [33], rational Jacobi [32], rational Euler [35], etc. to solve some ODEs. A fractional order of the rational Gegenbauer (FRG) function is defined as follows:

$$FRG_n^a(L, \alpha, x) = G_n^a\left(\frac{x^\alpha - L}{x^\alpha + L}\right), \tag{12}$$

in which  $L$  and  $\alpha$  are real positive numbers. FRG functions are orthogonal functions in semi-infinite intervals same as Eq. (9) according to the weight function  $w(x) = (1 - (\frac{x^\alpha - L}{x^\alpha + L}))^{a-\frac{1}{2}} \frac{2\alpha L x^{\alpha-1}}{(x^\alpha + L)^2}$  :

$$\int_0^\infty FRG_n^a(L, \alpha, x)FRG_m^a(L, \alpha, x)w(x)dx = \frac{\pi 2^{1-2a}\Gamma(n+2a)}{n!(n+a)(\Gamma(a))^2} \delta_{nm}. \tag{13}$$

For more information about fractional order functions, interested readers can see [38–41].

### 3.3. Approximation of functions

**Definition 1** Consider  $\Gamma = \{x|0 \leq x \leq \infty\}$  and  $L_w^2(\Gamma) = \{f : \Gamma \rightarrow \mathbb{R} | f \text{ is measurable and } \|f\|_w < \infty\}$  where,

$$w(x) = (1 - (\frac{x^\alpha - L}{x^\alpha + L}))^{a-\frac{1}{2}} \frac{2\alpha L x^{\alpha-1}}{(x^\alpha + L)^2},$$

and

$$\|f(x)\|_w = \left( \int_0^\infty f^2(x)w(x)dx \right)^{\frac{1}{2}},$$

is the norm induced by the inner product of the space

$$\langle f(x), g(x) \rangle_w = \int_0^\infty f(x)g(x)w(x)dx.$$

Any function  $y(x) \in C(0, \infty)$  can be expanded as the follows:

$$y(x) = \sum_{n=0}^\infty a_n FRG_n^a(L, \alpha, x), \tag{14}$$

where

$$a_i = \langle y(x), FRG_i^a(L, \alpha, x) \rangle = \left\langle \sum_{n=0}^\infty a_n FRG_n^a(L, \alpha, x), FRG_i^a(L, \alpha, x) \right\rangle, \tag{15}$$

that is,

$$a_n = \frac{n!(n+a)(\Gamma(a))^2}{\pi 2^{1-2a}\Gamma(n+2a)} \int_0^\infty FRG_n^a(L, \alpha, x)y(x)w(x)dx, \tag{16}$$

Now let assume

$$V_m = span\{FRG_0^a(L, \alpha, x), FRG_1^a(L, \alpha, x), \dots, FRG_m^a(L, \alpha, x)\},$$

is a finite dimensional subspace, therefore  $V_m$  is a complete subspace of  $L_w^2(\Gamma)$  [33, 42, 43]. Let define the  $L_w^2(\Gamma)$ -orthogonal projection  $\Pi_{N,w} : L_w^2(\Gamma) \rightarrow V_m$ , that for any function  $y \in L_w^2(\Gamma)$ :

$$\langle \Pi_{N,w}y - y, v \rangle = 0, \quad \forall v \in V_m. \tag{17}$$

It is clear that  $\Pi_{N,w}y$  is the best approximation of  $y(x)$  in  $V_m$  and can be expanded as [44]:

$$\Pi_{N,w}y = y_m(x) = \sum_{i=0}^m a_i FRG_i^a(L, \alpha, x). \tag{18}$$

## 4. Application of the methods

In this section, two approaches based on the Newton method and spectral collocation algorithm are explained to approximate the solution of the Thomas-Fermi equation. In one of them, we use the Newton method to linearize the Thomas-Fermi equation and then solve the several linear ODEs by a spectral method that we call this method pre-Newton method. In the other method, we convert the Thomas-Fermi equation to a nonlinear system of algebraic equations by using a spectral algorithm, then, solve this nonlinear system by using the classical Newton method which we call this method post-Newton. These two schemes are illustrated as follows.

### 4.1. Pre-Newton approach for Thomas-Fermi equation

Solving a system of nonlinear algebraic equations by using traditional Newton-type solvers have three major practical difficulties. The first one is selecting the start point which yields the convergence of the iterations. The second one is computing the Jacobian matrix of the system of equations at each iteration which has a lot of computational load on the algorithm. The last one is inverting a Jacobian matrix at each iteration which is the most expensive step of the algorithm. In the post-Newton approaches for solving nonlinear ODEs, we should overcome these difficulties [42]. In order to avoid these difficulties, we can apply the Newton method directly to the nonlinear ODE. In the next part, a famous Newton-type algorithm is described which converts nonlinear ODEs to a sequence of linear differential equations.

#### 4.1.1. Newton-Kantorovich method

Newton-Kantorovich method is a well-known and strong approach to converting nonlinear ODEs to linear ones which was introduced by Bellman and Kalaba [45-47]. This approach obtains the solution of a nonlinear ODE by solving a sequence of linear differential equations [48]. In fact, approximating the solution of a nonlinear equation is more complicated than a linear one; therefore, by using Newton-Kantorovich method, the solution of the sequence of the linear differential equations converges to the solution of the original nonlinear ODE [45, 49, 50]. This method is based on approximating a nonlinear function by using the linear part of the Taylor expansion of that function.

This fact can be extended to linearize a nonlinear ODE. In order to show how Newton-Kantorovich method works we consider a  $n$ -th order nonlinear ODE over the interval  $[0, b]$  as follows [50]:

$$L^{(n)}y(x) = f(y(x), y^{(1)}(x), \dots, y^{(n-1)}(x), x), \tag{19}$$

with the following boundary conditions:

$$B_k(y(0), y^{(1)}(0), \dots, y^{(n-1)}(0)) = 0 \quad k = 1, 2, \dots, l, \tag{20}$$

and

$$B_k(y(b), y^{(1)}(b), \dots, y^{(n-1)}(b)) = 0 \quad k = l + 1, l + 2, \dots, n, \tag{21}$$

where  $L^{(n)}$  is a linear  $n$ -th order ordinary differential operator and  $f$  and  $B_1, B_2, \dots, B_n$  are nonlinear functions of  $y(x)$  and its  $n - 1$  derivatives  $y^{(s)}, s = 1, 2, \dots, n - 1$ . If we apply Newton-Kantorovich method on Eq. (19) the  $(r + 1)$ -th iterative approximation of  $y(x)$  is obtained by solving follow linear ODE,

$$L^{(n)}y_{r+1}(x) = f(y_r(x), y_r^{(1)}(x), \dots, y_r^{(n-1)}(x), x) + \sum_{s=0}^n (y_{r+1}^{(s)}(x) - y_r^{(s)}(x)) f_{y^{(s)}}(y_r(x), y_r^{(1)}(x), \dots, y_r^{(n-1)}(x), x), \tag{22}$$

where  $y_r^0(x)$  is a notation for  $y_r(x)$ . Also, the linearized boundary conditions are obtained as follows:

$$\sum_{s=0}^{n-1} (y_{r+1}^{(s)}(0) - y_r^{(s)}(0)) B_{ky^{(s)}}(y_r(0), y_r^{(1)}(0), \dots, y_r^{(n-1)}(0), 0) = 0, \quad k = 1, \dots, l, \tag{23}$$

and

$$\sum_{s=0}^{n-1} (y_{r+1}^{(s)}(b) - y_r^{(s)}(b)) B_{ky^{(s)}}(y_r(b), y_r^{(1)}(b), \dots, y_r^{(n-1)}(v), b) = 0, \quad k = 1, \dots, l. \tag{24}$$

It is worth to mention that in the above formulas  $f_{y^{(s)}} = \frac{\partial f}{\partial y^{(s)}}$  and  $B_{ky^{(s)}} = \frac{\partial B_k}{\partial y^{(s)}}$  for  $s = 0, 1, \dots, n - 1$

By implementing Newton–Kantorovich method on Eq. (3), the  $(i + 1) - th$  iteration linear ODE for approximating the solution of Thomas-Fermi equation is as follows ( $i = 0, 1, 2, \dots$ ):

$$\sqrt{x}y_{i+1}''(x) - \frac{3}{2}(y_i(x))^{\frac{1}{2}}y_{i+1}(x) = \frac{-1}{2}(y_i(x))^{\frac{3}{2}}, \tag{25}$$

with the following boundary conditions:

$$y_{i+1}(0) = 1, \quad \lim_{x \rightarrow \infty} y_{i+1}(x) = 0. \tag{26}$$

An initial guess  $y_0(x)$  is required for the first step of the Newton–Kantorovich method. It is proved that when the initial guess satisfies one of the boundary conditions, the Newton–Kantorovich method will be convergent [45]. Thus, we consider  $y_0(x)=1$ .

#### 4.1.2. Collocation method in the pre-Newton method

The spectral collocation method based on FRG functions is applied to Eq. (25) at each iteration. According to the boundary conditions in Eq. (26), we approximate  $y_{i+1}(x)$  in  $(i + 1) - th$  iteration as:

$$y_{i+1}(x) \simeq y_{i+1}^N(x) = 1 + x \sum_{j=0}^{N-1} a_j^{i+1} FRG_j^a(L, \alpha, x). \tag{27}$$

where  $a_j^{i+1}$  is the  $j - th$  unknown coefficient in  $(i + 1) - th$  iteration. Equation (27) satisfies the boundary condition  $y(0) = 1$ . To satisfy the other boundary condition, we choose a sufficiently large number  $K$  and consider  $y_{i+1}(K) = 0$ . The Eq. (27) is replaced in Eq. (25); afterwards, the residual function is obtained:

$$Res_{i+1}(x) = \sqrt{x}y_{i+1}^{N''}(x) - \frac{3}{2}(y_i^N(x))^{\frac{1}{2}}y_{i+1}^N(x) + \frac{1}{2}(y_i^N(x))^{\frac{3}{2}}, \tag{28}$$

The roots of  $FRG_N^a(L, \alpha, x)$  are considered as the collocation points which are collocated in Eq. (28) and a system of linear algebraic equations is established. By solving this system at each iteration,  $y(x)$  is approximated.

$$Res_{i+1}(x_j) = 0, \quad j = 0, \dots, N - 1. \tag{29}$$

#### 4.2. Post-Newton approach for Thomas-Fermi equation

In the post-Newton approach for solving the Thomas-Fermi equation, we use a fully spectral technique same collocation method to solve the nonlinear equation Eq. (3) without any linearization method. In this method, by using the spectral collocation method based on the fractional order of rational Gegenbauer functions, we convert the nonlinear Thomas-Fermi equation Eq. (3) to a system of nonlinear algebraic equations. In this method, the unknown solution  $y(x)$  of Thomas-Fermi is approximated by the following series:

$$y(x) \simeq y_N(x) = 1 + x \sum_{j=0}^{N-1} a_j FRG_j^a(L, \alpha, x). \tag{30}$$

Then by substitution  $y^N(x)$  instead of  $y(x)$  in Eq. (3) the residual function is constructed as follows:

$$Res(x) = \sqrt{x}y^{N''}(x) - \frac{1}{\sqrt{x}}(y^N(x))^{\frac{3}{2}}. \tag{31}$$

Now, there are  $N$  unknown coefficients  $a_i, i = 0, 1, \dots, N - 1$ , to find these unknowns we need  $N$  equations. By using the collocation technique and roots of the fractional order of rational Gegenbauer function of order  $N$ , and by substitution of these nodes in the residual function we construct  $N$  nonlinear equations as follows:

$$F_i = Res(x_i) = 0, \quad i = 0, 1, \dots, N - 2,$$

to satisfy the boundary condition in infinite we set  $F_{N-1} = y_N(L) = 0$  for sufficient large  $L$ .

So,  $F : \mathbb{R}^N \rightarrow \mathbb{R}^N$  is a nonlinear function, and therefore, finding the solution of Eq. (3) has been transformed to find the solution of the nonlinear system of equations:

$$F(A) = 0, \quad A = [a_0, a_1, \dots, a_{N-1}]^T,$$

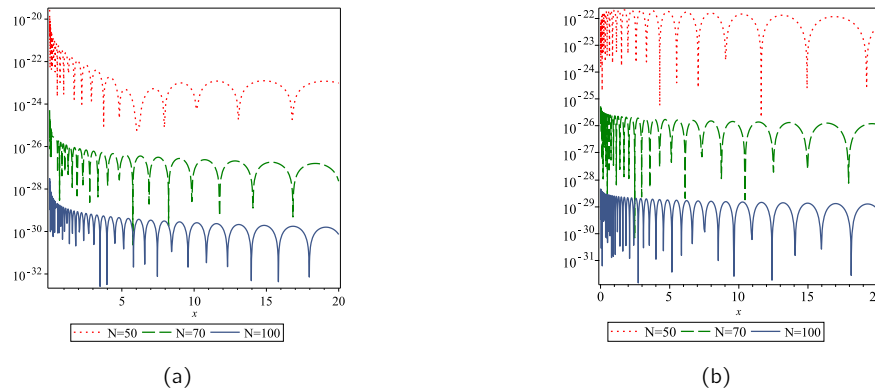


Figure 1. Graph of the logarithm of the absolute residual errors for different  $N$  for (a) pre-Newton and (b) post-Newton.

Now, we have transformed solving the nonlinear differential equation to finding the root of a nonlinear  $\mathbb{R}^N \rightarrow \mathbb{R}^N$  function.

One of the best methods to solve a nonlinear system is the classical Newton iterative method, that by using Tylor expansion:

$$F(x_{n+1}) = F(x_n) + (x_n - x_{n+1})F'(x_n), \tag{32}$$

presuppose  $x_{n+1}$  be root of  $F(x)$ ,  $F(x_{n+1}) = 0$  :

$$F(x_n) + (x_{n+1} - x_n)F'(x_n) = 0, \tag{33}$$

$$\Rightarrow x_{n+1} = x_n - F'(x_n)^{-1}F(x_n), \tag{34}$$

$F'(x) = J(x)$  is the  $n \times n$  Jacobian matrix and is defined as follows:

$$J_{ij} = \left( \frac{\partial f_i}{\partial x_j} \right), \tag{35}$$

therefore:

$$x_{n+1} = x_n - J(x_n)^{-1}F(x_n). \tag{36}$$

In fact, in each iteration, a linear system must be solved:

$$\begin{cases} x_{n+1} = x_n + \delta x_n \\ J(x_n)\delta x_n = F(x_n). \end{cases} \tag{37}$$

In this paper, we use  $LU$  method to solve linear system  $J(x_n)\delta x_n = F(x_n)$  in each iteration of Newton method. The initial guess of the post-Newton method is the simple vector  $x_0 = (1, 1, \dots, 1)^T$ .

### 5. Numerical results and discussion

According to Boyd’s book [42],  $L$  can be chosen by "The experimental trial-and-error method"; so, we consider  $L = 3$  in pre-Newton and  $L = 2.828$  in post-Newton, also we consider  $\alpha = \frac{1}{2}, a = \frac{1}{2}$  in the both and report the results. It is worth mentioning that all the computations are done by Maple, in a personal computer with the following hardware configuration: desktop 64-bit Intel Core i5 CPU, 8GB of RAM, 64-bit Operating System. In [37], Zhang and Boyd calculated an approximate solution for  $y'(0)$  with high accuracy; thus, the results of this study are compared with [37] and found that the obtained results are as accurate as [37]. The logarithm of absolute residual error for the Thomas-Fermi equation in the best iteration is represented in Fig. 1. This figure shows when the number of collocation points increases, the residual error tends to zero. The value of  $y'(0)$  is presented in Table 3 and compared with the obtained solution by state-of-the-art methods. Table 4 contains the values of  $y(x)$  and  $y'(x)$  for different values of  $x$ .

**Table 3.** Comparison of the obtained values of  $y'(0)$  by some researchers

Author/Authors	Year	Obtained value of $y'(0)$
Boyd [30]	(2013)	-1.5880710226113753127186845
Parand et al [32]	(2017)	-1.588071022611375312718684509423950109
Parand and Delkhosh (N=300) [33]	(2017)	-1.58807102261137531271868450942395010951
Zhang and Boyd (N=600) [37]	(2018)	-1.588071022611375312718684509423950109452746621674825616765677
pre-Newton (N=100)		-1.58807102261137531271868450942395010945274662
post-Newton (N=200)		-1.588071022611375312718684509423950109452746621674825616765677

**Table 5.** Runtime for the proposed methods with the different values of  $N$  and iteration

$N$	Iteration	Runtime for pre-Newton (s)	Runtime for post-Newton (s)
50	20	39.901	32.392
	30	60.626	48.688
	40	80.361	56.359
70	20	104.191	76.690
	30	172.292	111.274
	40	211.721	136.428
100	20	343.672	208.404
	30	469.262	304.123
	40	622.255	392.937

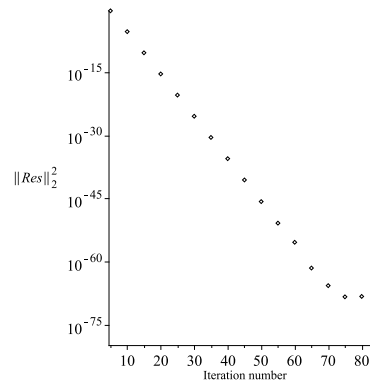


Figure 2. Graph of the  $\log(\|Res\|_2^2)$  for the post-Newton method with  $N = 200$  at different iterations.

**Table 4.** Values of  $y(x)$  and  $y'(x)$  obtained by the presented methods for the various values of  $x$

$y(x)$ and $y'(x)$	$x$	pre-Newton ( $N=100$ and iteration=40)	post-Newton ( $N=200$ and iteration=85)
$y(x)$	0.5	0.6069863833559799094944460701740221017049	0.6069863833559799094944460701740842378463
	3	0.1566326732164958413398134404775366125433	0.1566326732164958413398134404779118302783
	10	0.0243142929886808641901103881732913695553	0.0243142929886808641901103881763049683685
	50	0.0006322547829849047267797787287302055560	0.0006322547829849047267797787427886658114
	200	0.0000145018034969457646803986629623432665	0.0000145018034969457646803987687276929118
	5000	0.000000011309267063430848076021125559361	0.000000011309267063430848263855178787850
$y'(x)$	0.5	-0.4894116125745380886470058475611743123609	-0.4894116125745380886470058475573462887337
	3	-0.0624571308541209762287048999941581989893	-0.0624571308541209762287048999995217973789
	10	-0.0046028818712692545025435118554873081322	-0.0046028818712692545025435118515886154232
	50	-0.0000324989020482588146242006692476761611	-0.0000324989020482588146242006802396097650
	200	-0.0000002057532316475268926057043855114949	-0.0000002057532316475268926056858363001742
	5000	-0.000000000006753397121638834659796119395	-0.000000000006753397121638835144503744957

One of the advantages of the post-Newton approach is its computational speed. This approach is much faster than pre-Newton; as the iterations can be increased to 85 with an acceptable runtime. In Table 5, pre-Newton and post-Newton methods are compared in runtime with the different number of collocation points and iterations. It is derived that post-Newton is much faster than the other approach; therefore, we can consider a larger number of iterations for the post-Newton than pre-Newton. The logarithm of  $\|Res\|_2^2$  at different iterations of the post-Newton method for the Thomas-Fermi equation by using 200 points is represented in Fig. 2.

## 6. Conclusion

In this paper, we introduced and compared two points of view to solve nonlinear boundary problems over the semi-infinite interval. These two approaches are called the pre-Newton method and post-Newton method, respectively. The pre-Newton method is based on applying Newton–Kantorovich algorithm to the nonlinear ODE and solving the obtained linear ODEs from Newton–Kantorovich method by using the collocation algorithm. The post-Newton method is based on applying the collocation algorithm directly to the nonlinear ODE and then solving the obtained nonlinear system of algebraic equations by the classical iterative Newton method. The collocation algorithm which is used is based on orthogonal functions in the interval  $[0, \infty)$  which are called the fractional order of the rational Gegenbauer. Since the significance of the Thomas-Fermi equation, here, we consider



it as a test problem. In the Thomas–Fermi equation, the value of  $y'(0)$  has important information in physics and scientists attempt to approximate that precisely. Therefore, we compare the approximation solution in  $y'(0)$  with the other numerical methods and realize that our proposed method is effective. The approximate solutions for  $y(x)$  and  $y'(x)$  for various values of  $x$  are represented. Additionally, the suggested methods are compared in runtime to find out which method is more efficient. According to the results, the post-Newton approach is faster and more accurate than the pre-Newton approach. It is worth mentioning that one of the limitations of the proposed algorithms is the ill-posedness of systems of algebraic equations. This limitation causes we can not increase the number of collocation points. Moreover, the basis functions are too complex, and finding their roots as the collocation points is very time-consuming.

## References

1. G. Adomian, Solution of the Thomas–Fermi equation. *Applied Mathematics Letters*, 11:131–133, 1998. .
2. A. M. Wazwaz. The modified decomposition method and Padé approximants for solving the Thomas–Fermi equation. *Applied Mathematics and Computation*, 105:11–19, 1999.
3. R. Rach, J. S. Duan, A. M. Wazwaz. Solving coupled Lane–Emden boundary value problems in catalytic diffusion reactions by the Adomian decomposition method. *Journal of Mathematical Chemistry*, 52:255–267, 2014.
4. L. Epele, H. Fanchiotti, C. Canal, J. Ponciano. Padé approximate approach to the Thomas–Fermi problem. *Physical Review A*, 60:280–283, 1999.
5. B.J. Noye, M. Dehghan. New explicit finite difference schemes for two-dimensional diffusion subject to specification of mass. *Numerical Methods for Partial Differential Equations*, 15:521–534, 1999.
6. W. Bu, Y. Ting, Y. Wu, J. Yang. Finite difference/finite element method for two-dimensional space and time fractional Bloch–Torrey equations. *Journal of Computational Physics*, 293:264–279, 2015.
7. H. J. Choi, J. R. Kweon. A finite element method for singular solutions of the Navier–Stokes equations on a non-convex polygon. *Journal of Computational and Applied Mathematics*, 292:342–362, 2016.
8. F. Bayatbabolghani, K. Parand. Using Hermite function for solving Thomas–Fermi equation. *International Journal of Mathematics and Computer in Engineering*, 8:123–126, 2014.
9. K. Parand, M. Hemami. Numerical study of astrophysics equations by meshless collocation method based on compactly supported radial basis function. *International Journal of Applied and Computational Mathematics*, 3:1053–1075, 2016.
10. K. Parand, S. Abbasbandy, S. Kazem, A. R. Rezaei. An improved numerical method for a class of astrophysics problems based on radial basis functions. *Physica Scripta*, 83(1):015011, 2011.
11. S. Kazem, J. A. Rad, K. Parand, M. Shaban, H. Saber. The numerical study on the unsteady flow of gas in a semi-infinite porous medium using an RBF collocation method. *International Journal of Computer Mathematics*, 89(16):2240–2258, 2012.
12. S. Zhu, H. Zhu, Q. Wu, Y. Khan. An adaptive algorithm for the Thomas–Fermi equation. *Numerical Algorithms*, 59 (3):359–372, 2012.
13. A. Cedillo. A perturbative approach to the Thomas–Fermi equation in terms of the density. *Journal of Mathematical Physics*, 34:27132717, 1993.
14. K. Parand, A. Pirkhedri, M. Dehghan. The Sinc-collocation method for solving the Thomas–Fermi equation. *Journal of Computational and Applied Mathematics* 273 (2013) 244–252.
15. H.T. Davis. *Introduction to Nonlinear Differential and Integral Equations*. Dover, New York, 1962.
16. L.H. Thomas. The calculation of atomic fields. *Mathematical Proceedings of the Cambridge Philosophical Society*, 23:542–548, 1927.
17. E. Baker. The application of the Fermi–Thomas statistical model to the calculation of potential distribution in positive ions. *Quarterly of Applied Mathematics*, 36:630–647, 1930.
18. B. Laurenzi. An analytic solution to the Thomas–Fermi equation. *Journal of Mathematical Physics*, 10:2535–2537, 1990.
19. A. Sommerfeld. Asymptotische integration der differentialgleichung des Thomas–Fermischen atoms. *Zeitschrift für Physik*, 78:283–308, 1932.
20. J. He. Variational approach to the Thomas–Fermi equation. *Applied Mathematics and Computation*, 143:533–535, 2003.
21. J. Ramos. Piecewise quasilinearization techniques for singular boundary-value problems. *Computer Physics Communications*, 158:12–25, 2004.
22. N. Zaitsev, I. Matyushkin, D. Shamonov. Numerical solution of the Thomas–Fermi equation for the centrally symmetric atom. *Russian Microelectronics*, 33:303–309, 2004.
23. R. Iacono. An exact result for the Thomas–Fermi equation: a priori bounds for the potential slope at the origin. *Journal of Physics A: Mathematical and Theoretical*, 41:455204 7, 2008.
24. K. Parand, M. Shahini. Rational Chebyshev pseudospectral approach for solving Thomas–Fermi equation. *Physics Letters A*, 373:210–213, 2009.
25. A. Ebaid. A new analytical and numerical treatment for singular two-point boundary value problems via the Adomian decomposition method. *Journal of Computational and Applied Mathematics*, 235:1914–1924, 2011.
26. V. Marinca, N. Herisanu. An optimal iteration method with application to the Thomas–Fermi equation. *Central European Journal of Physics*, 9:891–895, 2011.
27. S. Abbasbandy, C. Bervillier. Analytic continuation of Taylor series and the boundary value problems of some nonlinear ordinary differential equations. *Applied Mathematics and Computation*, 218:2178–2199, 2011.
28. F. Fernandez. Rational approximation to the Thomas–Fermi equations. *Applied Mathematics and Computation*, 217:6433–6436, 2011.
29. M. Türkyilmazoglu. Solution of the Thomas–Fermi equation with a convergent approach. *Communications in Nonlinear Science and Numerical Simulation*, 17:4097–4103, 2012.
30. J. Boyd. Rational Chebyshev series for the Thomas–Fermi function: Endpoint singularities and spectral methods. *Journal of Computational and Applied Mathematics*, 244:90–101, 2013.

31. A. Kilicman, I. Hashimb, M. Tavassoli Kajani, M. Maleki. On the rational second kind Chebyshev pseudospectral method for the solution of the Thomas–Fermi equation over an infinite interval. *Journal of Computational and Applied Mathematics*, 257:79–85, 2014.
32. K. Parand, P. Mazaheri, H. Yousefi, M. Delkhosh. Fractional order of rational Jacobi functions for solving the non-linear singular Thomas-Fermi equation. *The European Physical Journal Plus*, 132(77), 2017.
33. K. Parand, M. Delkhosh. Accurate solution of the Thomas–Fermi equation using the fractional order of rational Chebyshev functions. *Journal of Computational and Applied Mathematics*, 317:624–642, 2017.
34. K. Parand, A. Ghaderi, H. Yousefi, M. Delkhosh. A new approach for solving nonlinear Thomas–Fermi equation based on fractional order of rational Bessel functions. *Electronic Journal of Differential Equations*, 2016(331), 2016.
35. K. Parand, H. Yousefi, M. Delkhosh, A. Ghaderi. A novel numerical technique to obtain an accurate solution to the Thomas–Fermi equation. *The European Physical Journal Plus*, 131(228), 2016.
36. Z. Sabir, M. A. Manzar, M. A. Zahoor Raja, M. Sheraz and A. M. Wazwaz. Neuro-heuristics for nonlinear singular Thomas-Fermi systems. *Applied Soft Computing*, 65:52–169, 2018.
37. X. Zhang, J. Boyd. Revisiting the Thomas-Fermi equation: Accelerating rational Chebyshev series through coordinate transformations. *Applied Numerical Mathematics*, 135:186–205, 2019.
38. A. H. Hadian Rasanan, S. Nedaei Janbesaraei, D. Baleanu. Fractional Chebyshev Kernel Functions: Theory and Application. *In Learning with Fractional Orthogonal Kernel Classifiers in Support Vector Machines: Theory, Algorithms and Applications*. Singapore: Springer Nature Singapore, 39–68, 2023.
39. A. Azmoon, S. Chakraverty, S. Kumar. Fractional Legendre Kernel Functions: Theory and Application. *In Learning with Fractional Orthogonal Kernel Classifiers in Support Vector Machines: Theory, Algorithms and Applications*. Singapore: Springer Nature Singapore, 69–91, 2023.
40. S. Nedaei Janbesaraei, A. Azmoon, D. Baleanu. Fractional Gegenbauer Kernel Functions: Theory and Application. *In Learning with Fractional Orthogonal Kernel Classifiers in Support Vector Machines: Theory, Algorithms and Applications*. Singapore: Springer Nature Singapore, 93–118, 2023.
41. A. H. Hadian Rasanan, J. A. Rad, M. S. Tameh, A. Atangana. Fractional Jacobi Kernel Functions: Theory and Application. *In Learning with Fractional Orthogonal Kernel Classifiers in Support Vector Machines: Theory, Algorithms and Applications*. Singapore: Springer Nature Singapore, 119–144, 2023.
42. J. P. Boyd. Chebyshev and Fourier Spectral Methods. *Courier Corporation*, 2001.
43. L. Fox, I.B. Parker. *Chebyshev Polynomials in Numerical Analysis*. Oxford university press, London, Vol. 29, 1968.
44. B. Y. Guo. *Spectral methods and their applications*. World Scientific, 1998.
45. R. Bellman, R. Kalaba. *Quasilinearization and Nonlinear Boundary-Value Problems*. Elsevier, New York, 1965.
46. S.D. Conte, C. de Boor. *Elementary Numerical Analysis*. McGraw-Hill International Editions, 1981.
47. A. Ralston, P. Rabinowitz. *A First Course in Numerical Analysis*. McGraw-Hill International Editions, 1988.
48. K. Parand, M. M. Moayeri, S. Latifi, M. Delkhosh. A numerical investigation of the boundary layer flow of an Eyring-Powell fluid over a stretching sheet via rational Chebyshev functions. *The European Physical Journal Plus*, 132:352, 2017.
49. V.B. Mandelzweig. Quasilinearization method and its verification on exactly solvable models in quantum mechanics. *Journal of Mathematical Physics*, 40:6266–6291, 1999.
50. V.B. Mandelzweig, F. Tabakin. Quasilinearization approach to nonlinear problems in physics with application to nonlinear ODEs. *Computer Physics Communications*, 141:268–281, 2001.
51. S. Yüzbaşı. Numerical solution of the Bagley–Torvik equation by the Bessel collocation method. *Mathematical Methods in the Applied Sciences*, 36:300–312, 2013.
52. K. Diethelm. *The Analysis of Fractional Differential Equations*. Springer-Verlag Berlin Heidelberg, Berlin, 2010.
53. S. Yüzbaşı. A numerical approximation for Volterra’s population growth model with fractional order. *Applied Mathematical Modelling*, 37:3216–3227, 2013.