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Faculty of Graduate Studies

**Linear Fredholm Integro-Differential Equation
of the Second Kind**

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Dedication

I dedicate this thesis to my beloved Palestine, my parents, my love my husband Ayman, my children Sandra and Kareem, my brother Khaled and my sisters Amany and Farah. Without their patience, understanding, support and most of all love, this work would not have been possible.

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الإقرار

أنا الموقع أدناه ، مقدم الرسالة التي تحمل العنوان :

Linear Fredholm Integro-Differential Equation of the Second Kind

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Declaration

The work provided in this thesis unless otherwise referenced, is the
researcher's own work , and has not been submitted elsewhere for any
other degree or qualification .

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Linear Fredholm Integro-Differential Equation of the Second Kind**By****Khulood Nedal Iseed Thaher****Supervisor****Prof. Naji Qatanani****Abstract**

In this thesis we focus on solving linear Fredholm integro-differential equation of the second kind due to its wide range of physical applications. We will investigate some analytical and numerical methods to solve this equation. The discussed analytical methods include: Direct computation method, variational iteration method, Adomian decomposition method, modified decomposition method, noise terms phenomenon and series solution method.

The numerical methods that will be presented here are: B-spline scaling function and wavelet method, Homotopy perturbation method, Legendre polynomial method and Taylor collocation method. Particular numerical examples demonstrating these numerical methods have been implemented for solving linear Fredholm integro-differential of the second kind. A comparison has been drawn between these numerical methods. Our numerical results show that the Homotopy perturbation method and Legendre polynomial method have proved to be the most efficient in comparison to the other numerical methods regarding their performance on the used examples.

Introduction

The subject of integro-differential equations is one of the most important mathematical tools in both pure and applied mathematics. Integro-differential equations play a very important role in modern science and technology such as heat transfer, diffusion processes, neutron diffusion and biological species. More details about the sources where these equations arise can be found in physics, biology and engineering applications as well as in advanced integral equations books. (see [2, 4, 5, 8, 18, 19, 20, 22]).

Some valid numerical methods, for solving Fredholm integro-differential equations have been developed by many researchers. In [7] Behiry and Hashish used wavelet methods for the numerical solution of Fredholm integro-differential equation. Lakestani, Razzaghi and Dehghan [25] applied linear semiorthogonal B-spline wavelets, specially constructed for the bounded interval to solve linear Fredholm integro-differential equation. In [16] Ji-Huan He solved linear Fredholm integro-differential equation by a Homotopy perturbation method. This method yields a very rapid convergence of the solution series in most cases, usually only few iterations leading to very accurate solutions. A Legendre collocation matrix method is presented to solve high-order linear Fredholm integro-differential equations under the mixed conditions in terms of Legendre polynomials were used in [42] by Yalcinbas, Sezer and Sorkan. In [23] Karamete and Sezer solved Fredholm integro-differential equation by a truncated Taylor series. Using the Taylor collocation points, this method transforms the

integro-differential equation to a matrix equation which corresponds to a system of linear algebraic equations with unknown Taylor coefficients.

However many approaches for solving the linear and nonlinear kind of these equations may be found in [10], [11], [13], [14], [15], [29], [30], [32], [36], [39] and [43].

In this work, some analytical methods have been used to solve the Fredholm integro-differential equation of the second kind. These methods are: direct computation method, variational iteration method, Adomian decomposition method, modified decomposition method, noise terms phenomenon and series solution method.

For the numerical treatment of the Fredholm integro-differential equation of the second kind, we have implemented the following methods: B-spline scaling functions and wavelets, Homotopy perturbation method, Legendre polynomial method and Taylor collocation method.

This thesis is organized as follow: In chapter one, we introduce some basic concepts of integro-differential equations. In chapter two, we investigate some analytical methods to solve the Fredholm integro-differential equation. These have been mentioned before. In chapter three, we implement some numerical methods for solving Fredholm integro-differential equation. These have mentioned before. Numerical examples and results are presented in chapter four and conclusions have been drawn.

Chapter One

Mathematical Preliminaries

Definition 1.1 [39]

An integro-differential equation is the equation in which the unknown function appears inside an integral sign and contains ordinary derivatives.

A standard integro-differential equation is of the form:

$$v^{(n)}(x) = f(x) + \lambda \int_{g(x)}^{h(x)} G(x, y) v(y) dy, \quad (1.1)$$

where $v^{(n)}(x) = \frac{d^n v}{dx^n}$, n integer, λ is a constant, $g(x)$ and $h(x)$ are limits of integration that may be both variables, constants, or mixed, $G(x, y)$ is a known function of two variables x and y called the kernel or the nucleus of the integro-differential equation. The function v that will be determined appears under the integral sign, and sometimes outside the integral sign. The function $f(x)$ and $G(x, y)$ are given.

1.1 Classification of Integro-Differential Equations

1.1.1 Types of integro-differential equations

There are three types of integro-differential equations:

1. Fredholm integro- differential equation

The Fredholm integro-differential equation of the second kind appears in the form:

$$v^{(n)}(x) = f(x) + \lambda \int_a^b G(x, y) v(y) dy. \quad (1.2)$$

2. Volterra integro-differential equation

The Volterra integro-differential equation of the second kind appears in the form:

$$v^{(n)}(x) = f(x) + \lambda \int_a^x G(x, y) v(y) dy, \quad (1.3)$$

where the upper limit of integration is variable.

3. Volterra-Fredholm integro-differential equation

The Volterra-Fredholm integro-differential equation of the second kind appear in two forms, namely:

$$v^{(n)}(x) = f(x) + \lambda_1 \int_a^x G_1(x, y) v(y) dy + \lambda_2 \int_a^b G_2(x, y) v(y) dy \quad (1.4)$$

and

$$v^{(n)}(x, y) = f(x, y) + \lambda \int_0^y \int_{\Omega} F(x, y, \xi, \tau, v(\xi, \tau)) d\xi d\tau, \quad x, y \in \Omega \times [0, Y] \quad (1.5)$$

where $f(x, y)$ and $F(x, y, \xi, \tau, v(\xi, \tau))$ are analytic functions,

$D = \Omega \times [0, Y]$ and Ω is a closed subset of R^n , $n = 1, 2, 3$. It is interesting to note that (1.4) contains disjoint Volterra and Fredholm integrals, whereas (1.5) contains mixed integrals. Other derivatives of less order may appear as well.

1.1.2 Linearity of integro-differential equation

Definition 1.2 [39]

The integro- differential equation

$$v^{(n)}(x) = f(x) + \lambda \int_{g(x)}^{h(x)} G(x, y) v(y) dy,$$

is said to be linear if the exponent of the unknown function $v(x)$ under the integral sign is one and the equation does not contain nonlinear functions of $v(x)$, otherwise, the equation is called nonlinear.

1.1.3 Homogeneity of integro-differential equation

Definition 1.3 [39]

The integro-differential equation

$$v^{(n)}(x) = f(x) + \lambda \int_{g(x)}^{h(x)} G(x, y) v(y) dy,$$

is said to be homogeneous if $f(x)$ is identically zero, otherwise it is called nonhomogeneous.

1.1.4 Singularity of integro-differential equation

When one or both limits of integration become infinite or when the kernel becomes infinite at one or more points within the range of integration. For example, the integro-differential equation.

$$v^{(n)}(x) = f(x) + \lambda \int_{-\infty}^{\infty} G(x, y) v(y) dy$$

is a singular integro-differential equation of the second kind.

(i) Singular integro-differential equation

If the kernel is of the form

$$G(x, y) = \frac{H(x, y)}{x - y} \quad (1.6)$$

where $H(x, y)$ is differentiable function $a \leq x \leq b$, $a \leq y \leq b$ with $H(x, y) \neq 0$, then the integro-differential equation is said to be a singular

equation with Cauchy kernel where the $G(x, y) = \int_a^b \frac{H(x, y)}{x - y} f(y) dy$ is

understood in the sense of Cauchy Principal value (CPV) and the notation

P.V. $\int_a^b \frac{H(x, y)}{x - y} dy$ is usually used to denote this. Thus

$$\text{P.V.} \int_a^b \frac{H(x, y)}{x - y} dy = \lim_{\varepsilon \rightarrow 0} \left\{ \int_a^{x-\varepsilon} \frac{H(x, y)}{x - y} dy + \int_{x+\varepsilon}^b \frac{H(x, y)}{x - y} dy \right\},$$

for example

$$v'(x) = \frac{2}{3}x^2 + 7x^4 + x^5 \ln\left(\frac{1-x}{1+x}\right) + \frac{2}{5} + \int_{-1}^1 \frac{v(y)}{x-y} dy, \quad |x| < 1.$$

(ii) Weakly singular integro-differential equation

The kernel is of the form

$$G(x, y) = \frac{H(x, y)}{|x - y|^\alpha} \quad (1.7)$$

where $H(x, y)$ is bounded *i.e.* several times continuously differentiable

$a \leq x \leq b$ and $a \leq y \leq b$ with $H(x, y) \neq 0$ and α is a constant such that

$0 < \alpha < 1$. For example, the equation

$$v^{(n)}(x) = \lambda \int_0^x \frac{1}{|x - y|^\alpha} v(y) dy, \quad 0 < \alpha < 1$$

is a singular integro-differential equation with a weakly singular kernel.

1.2 Systems of Fredholm Integro-Differential Equations

A system of Fredholm integro-differential equations of the second kind can be written as:

$$u^{(i)}(x) = f_1(x) + \int_a^b (G_1(x, y) u(y) + \tilde{G}_1(x, y) v(y)) dy \quad (1.8)$$

$$v^{(i)}(x) = f_2(x) + \int_a^b (G_2(x, y) u(y) + \tilde{G}_2(x, y) v(y)) dy.$$

The unknown functions $u(x)$ and $v(x)$, that will be determined, occur inside the integral sign whereas their derivatives appear mostly outside the

integral sign. The kernels $G_i(x, y)$, $\tilde{G}_i(x, y)$ and the functions $f_i(x)$ are given as real-valued functions [39].

1.3 Systems of Volterra Integro-Differential Equations

A system of Volterra integro-differential equation of the second kind has the form:

$$u^{(i)}(x) = f_1(x) + \int_0^x (G_1(x, y)u(y) + \tilde{G}_1(x, y)v(y)) dy \quad (1.9)$$

$$v^{(i)}(x) = f_2(x) + \int_0^x (G_2(x, y)u(y) + \tilde{G}_2(x, y)v(y)) dy.$$

The unknown functions $u(x)$ and $v(x)$ that will be determined, occur

inside the integral sign whereas their derivatives appear mostly outside the integral sign. The kernels $G_i(x, y)$, $\tilde{G}_i(x, y)$ and the functions $f_i(x)$ are given as real-valued functions [39].

1.4 Kinds of Kernels

1. Separable kernel

A kernel $G(x, y)$ is said to be separable or (degenerate) if it can be expressed in the form (1.10)

$$G(x, y) = \sum_{k=1}^n g_k(x)h_k(y),$$

where the functions $g_k(x)$ and the functions $h_k(y)$ are linearly independent. (see [39]).

2. Symmetric (or Hermitian) kernel

A complex-valued function $G(x, y)$ is called symmetric (or Hermitian) if

$$G(x, y) = G^*(x, y), \quad (1.11)$$

where the asterisk denotes the complex conjugate. For a real kernel, we have

$$G(x, y) = G(y, x). \quad (1.12)$$

Chapter Two

Analytical Methods for Solving Fredholm Integro-Differential Equation of the Second Kind

There are many analytical methods for solving Fredholm integro-differential equation of the second kind. In this chapter we will focus on the following methods: direct computation method, variational iteration method, Adomian decomposition method, modified decomposition method, noise terms phenomenon and series solution method.

2.1 Direct Computation Method

This method can be used to solve the Fredholm integro-differential equation of the second kind directly instead of a series form.

For the application of this method we consider the separable kernel of the form

$$G(x, y) = g(x) h(y), \quad (2.1)$$

We consider the Fredholm integro-differential equation of the general form

$$v^{(n)}(x) = f(x) + \int_a^b G(x, y) v(y) dy, \quad (2.2)$$

with the initial conditions $v^{(k)}(0) = b_k$, $0 \leq k \leq n - 1$.

Substituting (2.1) into (2.2) gives

$$v^{(n)}(x) = f(x) + g(x) \int_a^b h(y) v(y) dy. \quad (2.3)$$

Since the integral in equation (2.3) is a bounded integral and dependent only on one variable y , then we can assign this integral by a constant β ,

That is

$$\int_a^b h(y)v(y) dy = \beta. \quad (2.4)$$

Thus equation (2.3) becomes

$$v^{(n)}(x) = f(x) + \beta g(x). \quad (2.5)$$

Integrating both sides of (2.5) n times from 0 to x , also using the initial conditions, we can find formula for $v(x)$ that depends on β and x . This means we can write

$$v(x) = u(x, \beta). \quad (2.6)$$

Substituting (2.6) into the right side of (2.4), calculating the integral, also solving the resulting equation, we determine β . We obtain the exact solution $v(x)$ after substituting β into (2.6).

Example 2.1

Consider the Fredholm integro-differential equation

$$v'(x) = 12x + \int_0^1 v(y) dy \quad \text{with } v(0) = 0. \quad (2.7)$$

This equation may be written as

$$v'(x) = 12x + \beta, \quad v(0) = 0. \quad (2.8)$$

obtained by setting

$$\int_0^1 v(y) dy = \beta. \quad (2.9)$$

Integrating both sides of (2.8) from 0 to x , and by using the initial condition we obtain

$$v(x) = 6x^2 + \beta x. \quad (2.10)$$

Substituting (2.10) into (2.9) and evaluating the integral yield

$$\beta = \int_0^1 v(y) dy = 2 + \frac{1}{2}\beta \quad (2.11)$$

hence we find

$$\beta = 4. \quad (2.12)$$

The exact solution is given by

$$v(x) = 6x^2 + 4x. \quad (2.13)$$

2.2 Variational Iteration Method (VIM)

The variational iteration method (VIM) was established by Ji-Huan He [17]. The method provides rapidly convergent successive approximations of the exact solution only if such a closed form solution exists.

We consider the Fredholm integro-differential equation of the general form

$$v^{(i)}(x) = f(x) + \lambda \int_a^b G(x, y) v(y) dy. \quad (2.14)$$

The correction functional for this equation is given by

$$v_{n+1}(x) = v_n(x) + \int_0^x \lambda(\tau) [v_n^{(i)}(\tau) - f(\tau) - \int_a^b G(\tau, z) \tilde{v}_n(z) dz] d\tau. \quad (2.15)$$

The variational iteration method needs to determine the Lagrange multiplier $\lambda(\tau)$ via integration by parts and by using a restricted variation.

The variational iteration formula, without restricted variation, should be used for the determination of the successive approximations $v_{n+1}(x)$, $n \geq 0$ of the solution $v(x)$. The zeroth approximation v_0 can be any selective function. However, using the given initial values $v(0), v'(0), \dots$ are preferably used for the selective zeroth approximation v_0 .

Consequently, the solution is given by

$$v(x) = \lim_{n \rightarrow \infty} v_n(x). \quad (2.16)$$

Example 2.2

Consider the Fredholm integro-differential equation

$$v'(x) = 12x + \int_0^1 v(y) dy \quad \text{with} \quad v(0) = 0. \quad (2.17)$$

The correction functional for this equation is given by

$$v_{n+1}(x) = v_n(x) - \int_0^x [v_n'(\tau) - 12\tau - \int_0^1 v_n(z) dz] d\tau, \quad (2.18)$$

as $\lambda = -1$ for first-order Fredholm integro-differential equation.

This gives:

$$\begin{aligned}
 v_0(x) &= 0 \\
 v_1(x) &= v_0(x) - \int_0^x [v_0'(\tau) - 12\tau - \int_0^1 v_0(z) dz] d\tau = 6x^2 \\
 v_2(x) &= v_1(x) - \int_0^x [v_1'(\tau) - 12\tau - \int_0^1 v_1(z) dz] d\tau = 6x^2 + 2x \\
 v_3(x) &= v_2(x) - \int_0^x [v_2'(\tau) - 12\tau - \int_0^1 v_2(z) dz] d\tau = 6x^2 + 2x + x \quad (2.19) \\
 v_4(x) &= v_3(x) - \int_0^x [v_3'(\tau) - 12\tau - \int_0^1 v_3(z) dz] d\tau = 6x^2 + 2x + x + \frac{1}{2}x \\
 v_5(x) &= v_4(x) - \int_0^x [v_4'(\tau) - 12\tau - \int_0^1 v_4(z) dz] d\tau = 6x^2 + 2x + x + \frac{1}{2}x + \frac{1}{4}x \\
 &\vdots \\
 v_n(x) &= v_{n-1}(x) - \int_0^x [v_{n-1}'(\tau) - 12\tau - \int_0^1 v_{n-1}(z) dz] d\tau \\
 &= 6x^2 + 3x + \left(\frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots\right)x
 \end{aligned}$$

and so on. The VIM admits the use of

$$v(x) = \lim_{n \rightarrow \infty} v_n(x). \quad (2.20)$$

We obtain the exact solution

$$v(x) = 6x^2 + 4x. \quad (2.21)$$

2.3 Adomian Decomposition Method

The Adomian decomposition method (ADM) was introduced and developed by George Adomian [1-3]. ADM gives the solution in an infinite series of components. The idea of Adomian decomposition method is transforming Fredholm integro-differential equation to an integral equation.

We consider the second order Fredholm integro-differential equation of the second kind of the form .

$$v''(x) = f(x) + \int_a^b G(x, y) v(y) dy, \quad (2.22)$$

with the initial conditions $v(0) = b_0$, $v'(0) = b_1$.

Integrating both sides of (2.22) from 0 to x twice, we get

$$v(x) = b_0 + b_1x + L^{-1}(f(x)) + L^{-1}\left(\int_a^b G(x, y) v(y) dy\right) \quad (2.23)$$

where the initial conditions are used, and L^{-1} is a two-fold integral operator. $v(x)$ expressed as

$$v(x) = \sum_{n=0}^{\infty} v_n(x). \quad (2.24)$$

Setting (2.24) into (2.23) gives:

$$\sum_{n=0}^{\infty} v_n(x) = b_0 + b_1x + L^{-1}(f(x)) + L^{-1}\left(\int_0^x G(x, y) \sum_{n=0}^{\infty} v_n(y) dy\right) \quad (2.25)$$

or equivalently,

$$\begin{aligned}
 v_0(x) + v_1(x) + v_2(x) + \dots &= b_0 + b_1(x) + L^{-1}(f(x)) \\
 + L^{-1}\left(\int_0^x G(x,y)v_0(y)dy\right) &+ L^{-1}\left(\int_0^x G(x,y)v_1(y)dy\right) \\
 + L^{-1}\left(\int_0^x G(x,y)v_2(y)dy\right) &+ \dots
 \end{aligned} \tag{2.26}$$

Note that, $v_0(x)$ is defined by all the terms that are not included under the integral sign, that is

$$\begin{aligned}
 v_0(x) &= b_0 + b_1(x) + L^{-1}(f(x)) \\
 v_{k+1}(x) &= L^{-1}\left(\int_0^x G(x,y)v_k(y)dy\right), \quad k \geq 0.
 \end{aligned} \tag{2.27}$$

Using (2.24), the obtained series converges to the exact solution if such a solution exists, see ([12], [39]).

Example 2.3

Consider the Fredholm integro-differential equation

$$v'(x) = -1 + 24x + \int_0^1 v(y)dy \quad \text{with } v(0) = 0. \tag{2.28}$$

Note that, the integral at the right side is equivalent to a constant.

Integrating both sides of Eq. (2.28) from 0 to x , we get

$$v(x) - v(0) = -x + 12x^2 + x \left(\int_0^1 v(y) dy \right) \quad (2.29)$$

using the initial condition

$$v(x) = -x + 12x^2 + x \left(\int_0^1 v(y) dy \right). \quad (2.30)$$

Let

$$v(x) = \sum_{n=0}^{\infty} v_n(x). \quad (2.31)$$

Setting (2.31) into (2.30) gives

$$\sum_{n=0}^{\infty} v_n(x) = -x + 12x^2 + x \left(\int_0^1 v_n(y) dy \right). \quad (2.32)$$

The components v_j , $j \geq 0$ of $v(x)$ can be determined by using the recurrence relation

$$v_0(x) = -x + 12x^2, \quad v_{k+1}(x) = x \left(\int_0^1 v_k(y) dy \right), \quad k \geq 0. \quad (2.33)$$

This gives

$$\begin{aligned} v_0(x) &= -x + 12x^2, \quad v_1(x) = x \left(\int_0^1 v_0(y) dy \right) = \frac{7}{2}x, \\ v_2(x) &= x \left(\int_0^1 v_1(y) dy \right) = \frac{7}{4}x, \quad v_3(x) = x \left(\int_0^1 v_2(y) dy \right) = \frac{7}{8}x \end{aligned} \quad (2.34)$$

Hence, the solution is

$$v(x) = -x + 12x^2 + \frac{7}{2}x \left(1 + \frac{1}{2} + \frac{1}{4} + \dots\right), \quad (2.35)$$

which gives the exact solution

$$v(x) = 6x + 12x^2. \quad (2.36)$$

2.4 Modified Decomposition Method

As shown before, The Adomian decomposition method provides the solution in an infinite series of components. The method substitutes the decomposition series of $v(x)$, given by

$$v(x) = \sum_{n=0}^{\infty} v_n(x). \quad (2.37)$$

The process of converting Fredholm integro-differential equation to an integral equation can be achieved by integrating both sides of the Fredholm integro-differential equation from 0 to x as many times is needed to convert it into the integral equation

$$v(x) = f(x) + \lambda \int_a^b G(x, y) v(y) dy. \quad (2.38)$$

The standard Adomian decomposition method introduces the recurrence relation

$$v_0(x) = f(x) \quad (2.39)$$

$$v_{k+1}(x) = \lambda \int_a^b G(x, y) v_k(y) dy, \quad k \geq 0.$$

The modified decomposition method presents a slight variation to the recurrence relation (2.38) to determine the components of $v(x)$ in an easier and faster manner. For many cases, the function $f(x)$ can be set as the sum of two partial functions, namely $f_1(x)$ and $f_2(x)$. In other words

$$f(x) = f_1(x) + f_2(x). \quad (2.40)$$

The modified decomposition method admits the use of the modified recurrence relation

$$v_0(x) = f_1(x)$$

$$v_1(x) = f_2(x) + \lambda \int_a^b G(x, y) v_0(y) dy \quad (2.41)$$

$$v_{k+1}(x) = \lambda \int_a^b G(x, y) v_k(y) dy, \quad k \geq 1.$$

The difference between the recurrence relation (2.39) and the modified recurrence relation (2.41) is in the formation of $v_0(x)$ and $v_1(x)$ only. The other components $v_j(x)$, $j \geq 2$ remain the same in the two recurrence relations [39].

Example 2.4

Consider the Fredholm integro-differential equation

$$v''(x) = -\sin(x) + x - x \int_0^{\frac{\pi}{2}} y v(y) dy \quad \text{with } v(0) = 0, v'(0) = 1. \quad (2.42)$$

Using the modified decomposition method. First, integrate both sides of (2.42) from 0 to x twice, and using the given conditions, we obtain the Fredholm integral equation

$$v(x) = \sin(x) + \frac{1}{3!}x^3 - \frac{1}{3!}x^3 \int_0^{\frac{\pi}{2}} y v(y) dy. \quad (2.43)$$

$$f(x) = \sin(x) + \frac{1}{3!}x^3 \quad (2.44)$$

We split $f(x)$ into two parts, namely

$$f_1(x) = \sin(x)$$

$$f_2(x) = +\frac{1}{3!}x^3, \quad (2.45)$$

$f_1(x)$ related to the zeroth component, $v_0(x)$, and add $f_2(x)$ to the component $v_1(x)$. Therefore, we obtain the modified recurrence relation

$$v_0(x) = \sin(x)$$

$$v_1(x) = \frac{1}{3!}x^3 - \frac{1}{3!}x^3 \int_0^{\frac{\pi}{2}} y v_0(y) dy = 0, \quad (2.46)$$

all remaining components $v_2(x), v_3(x), \dots$ are zeros. This gives the exact solution as

$$v(x) = \sin(x) \quad (2.47)$$

2.5 The Noise Terms Phenomenon

The idea behind solving equations with noise terms if the noise terms appear between components of $v(x)$, then the exact solution can be obtained by considering only the first two components $v_0(x)$ and $v_1(x)$. The noise terms are defined as the identical terms, with opposite signs, that may appear between the components of the solution $v(x)$. The conclusion made by [1], [3], and [41] suggests that if we observe the appearance of identical terms in both components with opposite signs, then by canceling these terms, the remaining non-canceled terms of v_0 may in some cases provide the exact solution, that should be justified through substitution.

It was formally proved that other terms in other components will vanish in the limit if the noise terms occurred in $v_0(x)$ and $v_1(x)$.

Example 2.5

Consider the Fredholm integro-differential equation

$$v'(x) = 1 - \frac{1}{3}x + \int_0^1 xy v(y) dy \quad \text{with } v(0) = 0. \quad (2.48)$$

Using the noise terms phenomenon. By integrating both sides of the equation (2.48) from 0 to x gives

$$v(x) - v(0) = x - \frac{1}{6}x^2 + \frac{1}{2}x^2 \int_0^1 y v(y) dy, \quad v(0) = 0. \quad (2.49)$$

by using the initial condition

$$v(x) = x - \frac{1}{6}x^2 + \frac{1}{2}x^2 \int_0^1 y v(y) dy. \quad (2.50)$$

Let

$$v(x) = \sum_{n=0}^{\infty} v_n(x). \quad (2.51)$$

Substituting (2.51) into both sides of (2.50) yields

$$\sum_{n=0}^{\infty} v_n(x) = x - \frac{1}{6}x^2 + \frac{1}{2}x^2 \int_0^1 y \left(\sum_{n=0}^{\infty} v_n(y) \right) dy. \quad (2.52)$$

or equivalently

$$\begin{aligned} v_0(x) &= x - \frac{1}{6}x^2 \\ v_1(x) &= \frac{7}{48}x^2. \end{aligned} \quad (2.53)$$

Notice that $v_1(x)$ can be written as

$$v_1(x) = \frac{1}{6}x^2 - \frac{1}{48}x^2. \quad (2.54)$$

The noise terms $\pm \frac{1}{6}x^2$ appear between the components $v_0(x)$ and $v_1(x)$.

By canceling the noise term from $v_0(x)$, we obtain the exact solution

$$v(x) = x. \quad (2.55)$$

2.6 The Series Solution Method

The series method is useful method that stems mainly from the Taylor series for analytic functions for Fredholm integro-differential equation.

Definition 2.1 [39]

A real function $v(x)$ has the Taylor series representation

$$v(x) = \sum_{k=0}^n \frac{v^{(k)}(b)}{k!} (x-b)^k, \quad (2.56)$$

is called analytic if it has derivatives of all orders such that the Taylor series at any point b in its domain converges to $v(x)$ in a neighborhood of b .

For simplicity, the generic form of the Taylor series at $b=0$ is given as

$$v(x) = \sum_{n=0}^{\infty} a_n x^n. \quad (2.57)$$

We will assume that the Fredholm integro-differential equation of the second kind

$$v^{(k)}(x) = f(x) + \lambda \int_a^b G(x,y) v(y) dy, v_j(0) = a_j, 0 \leq j \leq (k-1) \quad (2.58)$$

is analytic, and therefore possesses a Taylor series of the form given in (2.57), where the coefficients a_n will be determined recurrently.

Setting (2.57) into (2.58) yields

$$\left(\sum_{n=0}^{\infty} a_n x^n \right)^{(k)} = T(f(x)) + \lambda \int_a^b G(x,y) \left(\sum_{n=0}^{\infty} a_n y^n \right) dy, \quad (2.59)$$

This is equivalent to

$$(a_0 + a_1x + a_2x^2 + \dots)^{(k)} = T(f(x)) + \lambda \int_a^b G(x, y) (a_0 + a_1y + a_2y^2 + \dots) dy \quad (2.60)$$

where $T(f(x))$ is the Taylor series for $f(x)$.

After performing integration in the right hand side of (2.60), we compare coefficients of power of x on both sides.

Example 2.6

Consider the Fredholm integro-differential equation of the second kind

$$v'(x) = 4x + \int_{-1}^1 (x - y)v(y) dy \quad \text{with } v(0) = 2, \quad (2.61)$$

using the series method.

Let

$$v'(x) = \left(\sum_{n=0}^{\infty} a_n x^n \right)'. \quad (2.62)$$

Inserting (2.62) into (2.61), yields

$$\sum_{n=0}^{\infty} n a_n x^{n-1} = 4x + \int_{-1}^1 ((x - y) \sum_{n=0}^{\infty} a_n y^n) dy. \quad (2.63)$$

$a_0 = 2$ from the given initial conditions. Evaluating the integral at the right side gives

$$\begin{aligned}
 a_1 + 2a_2x + 3a_3x^2 + \dots = \\
 -\frac{2}{3}a_1 - \frac{2}{5}a_3 - \frac{2}{7}a_5 - \frac{2}{9}a_7 - \dots + (4 + 2a_0 + \frac{2}{3}a_2 + \frac{2}{5}a_4 + \frac{2}{7}a_6 + \dots)x.
 \end{aligned} \tag{2.64}$$

Comparing coefficients of like powers of x in both sides of (2.64) gives

$$a_1 = 0, \quad a_2 = 6, \quad a_n = 0, \quad n \geq 2. \tag{2.65}$$

The exact solution is given by

$$v(x) = 2 + 6x^2. \tag{2.66}$$

where we used $a_0 = 2$ from the initial condition.

Chapter Three

Numerical Methods for Solving Fredholm

Integro-Differential Equation of the Second Kind

There are many numerical methods available for solving Fredholm integro-differential equation of the second kind . In this chapter we will discuss the following methods: B-spline scaling functions and wavelets, Homotopy perturbation method, Legendre polynomial method and Taylor collocation method.

3.1 B-Spline Scaling Functions and Wavelets on [0,1]

B-spline scaling functions and wavelets which are presented to approximate the solution of linear second order Fredholm integro-differential equation [6], [7] and [9]. Their properties and the operational matrices of derivative for these functions are presented to reduce the solution of linear Fredholm integro-differential equations to a solution of algebraic equations.

Consider the linear second-order Fredholm integro-differential equation of the form:

$$\sum_{i=0}^2 \tau_i(x) y^{(i)}(x) = f(x) + \int_0^1 K(x,t) y(t) dt, \quad 0 \leq x \leq 1 \quad (3.1)$$

with

$$y(0) = y_0, \quad y(1) = y_1 \quad (3.2)$$

where τ_i ($i = 0, 1, 2$), f , and K , are given functions in $L^2[0, 1]$, y_0 and y_1 are given real numbers and y is the unknown function to be found.

The second-order B-splines scaling functions are defined as:

$$\chi_{i,j}(x) = \begin{cases} x_i - j, & j \leq x_i < j + 1 \\ 2 - (x_i - j), & j + 1 \leq x_i < j + 2, j = 0, \dots, 2^i - 2 \\ 0, & \text{otherwise} \end{cases} \quad (3.4)$$

with the respective left- and right-hand side boundary scaling functions

$$\chi_{i,j}(x) = \begin{cases} 2 - (x_i - j), & 0 \leq x_i < 1, j = -1 \\ 0, & \text{otherwise} \end{cases} \quad (3.5)$$

$$\chi_{i,j}(x) = \begin{cases} x_i - j, & j \leq x_i < j + 1, j = 2^i - 1 \\ 0, & \text{otherwise} \end{cases} \quad (3.6)$$

the actual coordinate position x is related to x_i according to $x_i = 2^i x$.

The second-order B-spline wavelets are given by

$$\mu_{i,j}(x) = \frac{1}{6} \begin{cases} x_i - j, & j \leq x_i < j + 1/2 \\ 4 - 7(x_i - j), & j + 1/2 \leq x_i < j + 1 \\ -19 + 16(x_i - j), & j + 1 \leq x_i < j + 3/2 \\ 29 - 16(x_i - j), & j + 3/2 \leq x_i < j + 2 \\ -17 + 7(x_i - j), & j + 2 \leq x_i < j + 5/2 \\ 3 - (x_i - j), & j + 5/2 \leq x_i < j + 3 \\ 0, & \text{otherwise} \end{cases} \quad (3.7)$$

with the respective left- and right-hand side boundary wavelets:

$$\mu_{i,j}(x) = \frac{1}{6} \begin{cases} -6 + 23x_i, & 0 \leq x_i < 1/2 \\ 14 - 17x_i, & 1/2 \leq x_i < 1 \\ -10 + 7x_i, & 1 \leq x_i < 3/2 \\ 2 - x_i, & j + 3/2 \leq x_i < 2, j = -1 \\ 0, & \textit{otherwise} \end{cases} \quad (3.8)$$

$$\mu_{i,j}(x) = \frac{1}{6} \begin{cases} 2 - (j + 2 - x_i), & j \leq x_i < j + 1/2 \\ -10 + 7(j + 2 - x_i), & j + 1/2 \leq x_i < j + 1 \\ 14 - 17(j + 2 - x_i), & j + 1 \leq x_i < j + 3/2, j = 2^i - 2 \\ -6 + 23(j + 2 - x_i), & j + 3/2 \leq x_i < j + 2 \\ 0, & \textit{otherwise.} \end{cases} \quad (3.9)$$

From (3.4)-(3.9), we get more clear description for these two sets of equations

$$\chi_{i,j}(x) = \frac{1}{2} \chi_{i+1,2j}(x) + \chi_{i+1,2j+1}(x) + \frac{1}{2} \chi_{i+1,2j+2}(x), \quad i = 2, 3, \dots$$

$$j = 0, 1, 2, \dots, 2^j - 2$$

$$\chi_{i,-1}(x) = \frac{1}{2} \chi_{i+1,-1}(x) + \frac{1}{2} \chi_{i+1,0}(x), \quad i = 2, 3, \dots \quad (3.10)$$

$$\chi_{i,2^i-1}(x) = \chi_{i+1,2^{i+1}-1}(x) + \frac{1}{2} \chi_{i+1,2^i}(x), \quad i = 2, 3, \dots$$

and

$$\begin{aligned} \mu_{i,j}(x) = & \frac{1}{12} \chi_{i+1,2j}(x) - \frac{1}{2} \chi_{i+1,2j+1}(x) + \frac{5}{6} \chi_{i+1,2j+2}(x) \\ & - \frac{1}{2} \chi_{i+1,2j+3}(x) + \frac{1}{12} \chi_{i+1,2j+4}(x), \quad i = 2, 3, \dots, j = 0, 1, 2, \dots, 2^j - 2 \end{aligned}$$

$$\begin{aligned}\mu_{i,-1}(x) &= -\chi_{i+1,-1}(x) + \frac{11}{12}\chi_{i+1,0}(x) - \frac{1}{2}\chi_{i+1,1}(x) + \frac{1}{12}\chi_{i+1,2}(x), \\ \mu_{i,2^i-1}(x) &= -\chi_{i+1,2^{i+1}-2}(x) + \frac{11}{12}\chi_{i+1,2^{i+1}-3}(x) - \frac{1}{2}\chi_{i+1,2^{i+1}-4}(x) + \frac{1}{12}\chi_{i+1,2^{i+1}-5}(x).\end{aligned}\tag{3.11}$$

We define

$$\mathbf{X}_{M+1,k}(x) (k = -1, \dots, 2^{M+1} - 1) = \left[\chi_{M+1,-1}, \chi_{M+1,0}, \dots, \chi_{M+1,2^{M+1}-1} \right]^T \tag{3.12}$$

let

$$\mathcal{Q}_{M+1} = \int_0^1 \mathbf{X}_{M+1} \mathbf{X}_{M+1}^T dx. \tag{3.13}$$

The entry $(\mathcal{Q}_{M+1})_{ij}$ of the matrix \mathcal{Q}_{M+1} in (3.13) is calculated from

$$\int_0^1 \chi_{M+1,i}(x) \chi_{M+1,j}(x) dx \tag{3.14}$$

using (3.4)-(3.6), (3.12) and (3.14), we get a symmetric

$(2^{M+1} + 1) \times (2^{M+1} + 1)$ matrix for \mathcal{Q}_{M+1} which is given by:

$$\mathcal{Q}_{M+1} = \frac{1}{2^{M-1}} \begin{bmatrix} \frac{1}{12} & \frac{1}{24} & 0 & \dots & 0 \\ \frac{1}{24} & \frac{1}{6} & \frac{1}{24} & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & \frac{1}{24} & \frac{1}{6} & \frac{1}{24} \\ 0 & \dots & 0 & \frac{1}{24} & \frac{1}{12} \end{bmatrix}. \tag{3.15}$$

Finally, we want to define a vector M of order $(2^{(M+1)} + 1) \times 1$ as:

$$M = [\chi_{2,-1}, \chi_{2,0}, \dots, \chi_{2,3}, \mu_{2,-1}, \dots, \mu_{2,2}, \mu_{3,-1}, \dots, \mu_{3,6}, \dots, \mu_{M,-1}, \dots, \mu_{M,2^M-2}]^T. \quad (3.16)$$

The operational matrix of derivative

The differentiation of the vector X_{M+1} and M in (3.16) can be expressed as:

$$X'_{M+1} = D_X X_{M+1}, \quad M' = D_M M \quad (3.17)$$

where D_X and D_M are $(2^{M+1} + 2) \times (2^{M+1} + 2)$ operational matrices of derivative of B-spline scaling functions and wavelets respectively. The matrix D_X

$$D_X = \int_0^1 X'_{M+1}(t) \tilde{X}_{M+1}^T(t) dt = \left(\int_0^1 X'_{M+1}(t) X_{M+1}^T(t) dt \right) (Q_{M+1})^{-1} = R (Q_{M+1})^{-1} \quad (3.18)$$

where

$$R = \int_0^1 X'_{M+1}(t) X_{M+1}^T(t) dt. \quad (3.19)$$

In (3.19), R is $(2^{M+1} + 1) \times (2^{M+1} + 1)$ matrix given by

$$R = \begin{bmatrix} \int_0^1 \chi'_{M+1,-2}(t) \chi_{M+1,-2}(t) dt & \cdots & \int_0^1 \chi'_{M+1,-2}(t) \chi_{M+1,2^M-1}(t) dt \\ \vdots & \ddots & \vdots \\ \int_0^1 \chi'_{M+1,2^{M+1}-1}(t) \chi_{M+1,-2}(t) dt & \cdots & \int_0^1 \chi'_{M+1,2^{M+1}-1}(t) \chi_{M+1,2^{M+1}-1}(t) dt \end{bmatrix} \quad (3.20)$$

since the elements in the vector given in (3.12) are nonzero between

$k/2^{M+1}$ and $(k+2)/2^{M+1}$, then for any entries of $R_{j,k}$, we have

$$\left. \begin{aligned} R_{j,k} &= \int_0^1 \chi'_{M+1,j}(t) \chi_{M+1,k}(t) dt = \int_{k/2^{M+1}}^{(k+2)/2^{M+1}} \chi'_{M+1,j}(t) \chi_{M+1,k}(t) dt \\ &= \int_{k/2^{M+1}}^{(k+1)/2^{M+1}} \chi'_{M+1,j}(t) \chi_{M+1,k}(t) dt + \int_{(k+1)/2^{M+1}}^{(k+2)/2^{M+1}} \chi'_{M+1,j}(t) \chi_{M+1,k}(t) dt \\ &= \int_{k/2^{M+1}}^{(k+1)/2^{M+1}} 2^{M+1} \chi_{M+1,k}(t) dt + \int_{(k+1)/2^{M+1}}^{(k+2)/2^{M+1}} -2^{M+1} \chi_{M+1,k}(t) dt \end{aligned} \right\} \cdot (3.21)$$

From (3.21), we get

$$R = \begin{bmatrix} -\frac{1}{2} & -\frac{1}{2} & & & & \\ & \frac{1}{2} & 0 & -\frac{1}{2} & & \\ & & \ddots & \ddots & \ddots & \\ & & & \frac{1}{2} & 0 & -\frac{1}{2} \\ & & & & \frac{1}{2} & \frac{1}{2} \end{bmatrix}. \quad (3.22)$$

(3.32)

$$M' = G X'_{M+1} = G D_X X_{M+1} = G R (Q_{M+1})^{-1} X_{M+1} = G R (Q_{M+1})^{-1} G^{-1} M = D_M M,$$

where

$$D_M = G R (Q_{M+1})^{-1} G^{-1}. \quad (3.33)$$

we want to explain the technique. We solve linear Fredholm integro-differential equation by using B-spline wavelets. For this purpose we write (3.1) as

$$\sum_{i=0}^2 \tau_i(x) y^{(i)}(x) = f(x) + z(x) \quad 0 \leq x \leq 1, \quad (3.34)$$

where

$$z(x) = \int_0^1 K(x,t) y(t) dt. \quad (3.35)$$

We define

$$y(x) = C^T M(x) \quad (3.36)$$

$$z(x) = \int_0^1 K(x,t) C^T M(t) dt, \quad (3.37)$$

where $M(x)$ is defined in (3.16), and C is $(2^{M+1} + 1) \times 1$ unknown vector defined as

$$\left[c_{-1}, c_0, \dots, c_3, d_{2,-1}, \dots, d_{2,2}, d_{3,1}, \dots, d_{3,6}, \dots, d_{M,-1}, \dots, d_{M,2^M-2} \right]^T$$

We can approximate (3.37), using Newton-Cotes techniques as

$$z(x) = \int_0^1 K(x,t) C^T M(t) dt = \sum_{i=1}^n w_i K(x,t_i) C^T M(t_i) = F(x, C), \quad (3.38)$$

where w_i and t_i are weight and nodes of Newton-Cotes integration techniques respectively. Using (3.32) and (3.36), we get

$$\begin{aligned} y'(x) &= C^T M'(x) = C^T D_M M(x), \\ y''(x) &= C^T D_M'' M(x), \end{aligned} \quad (3.39)$$

From (3.30), (3.34), and (3.35), we get

$$\tau_0(x)C^T M(x) + \tau_1(x)C^T D_w M(x) + \tau_2(x)D_w^2 M(x) = f(x) + F(x, C). \quad (3.40)$$

Also, using (3.2) and (3.36), we have

$$C^T \Psi(0) = y_0, \quad C^T \Psi(1) = y_1. \quad (3.41)$$

To find the solution $y(x)$, we first collocate (3.40) in

$x_i = (2i - 1)/(2^{M+2} - 2)$, $i = 1, 2, \dots, 2^{M+1} - 1$, the resulting equation gener-

ates $2^{M+1} - 1$ linear equations which can be solved using Newton's

iterative method. The initial values required to start Newton's method have

been chosen by taking $y(x)$ as linear function between $y(0) = y_0$ and

$y(1) = y_1$. The total unknown for vector C is $2^{M+1} + 1$. These can be

obtained by using (3.40) and (3.41).

3.2 Homotopy Perturbation Method (HPM)

The Homotopy perturbation method was introduced and developed by Jihvan He [16]. The Homotopy perturbation method couples a Homotopy technique of topology and a perturbation technique.

Consider the following Fredholm integro-differential equation of the second kind of the form

$$v'(x) = f(x) + \lambda \int_0^1 G(x, y) v(y) dy. \quad (3.50)$$

We define the operator

$$L(u) = u'(x) - \lambda \int_0^1 G(x, y) u(y) dy - f(x) = 0 \quad (3.51)$$

where $u'(x) = v'(x)$. Next we define the Homotopy $H(u, m)$, $m \in [0, 1]$ by

$$H(u, 0) = F(u) \quad , \quad H(u, 1) = L(u), \quad (3.52)$$

where $F(u)$ is a functional operator. We construct a convex Homotopy of the form

$$H(u, m) = (1 - m)F(u) + mL(u). \quad (3.53)$$

This Homotopy satisfies (3.52) for $m = 0$ and $m = 1$ respectively. The embedding parameter m monotonously increases from zero to unity as the trivial problem $F(u) = 0$ is continuously deformed to the original problem $L(u) = 0$.

HPM uses the Homotopy parameter m as an expanding parameter to obtain

$$u = w_0 + mw_1 + m^2w_2 + m^3w_3 + \dots \quad (3.54)$$

when $m \rightarrow 1$, (3.54) corresponds to (3.53) and becomes the approximate solution of (3.51), i.e.,

$$v = \lim_{m \rightarrow 1} u = w_0 + w_1 + w_2 + w_3 + \dots \quad (3.55)$$

The series (3.55) is convergent for most cases and the rate of convergence depends on $L(u)$.

Assume $F(u) = u(x) - f(x)$, and substituting (3.54) in (3.51) and equating the terms with identical power of m , we have

$$m^0 : w_0'(x) = f(x) \quad (3.56)$$

$$m^n : w_n'(x) = \int_0^1 G(x, y) w_{n-1}(y) dy, n = 1, 2, \dots \quad (3.57)$$

Notice that the recurrence relation (3.57) is the same standard Adomian decomposition method as presented before in chapter two.

3.3 Legendre Polynomial Method

Orthogonal polynomials play a very important role in applications of mathematics, mathematical physics, engineering and computer science. One of the most common set of the Legendre polynomials $\{P_0(x), P_1(x), \dots, P_N(x)\}$ which are orthogonal on $[-1, 1]$ and satisfy the Legendre differential equation

$$(1-x^2)y''(x) - 2xy'(x) + n(n+1)y(x) = 0, \quad -1 < x < 1, \quad n \geq 0$$

and are given by the form

$$P_n(x) = \frac{1}{2^n} \sum_{k=0}^{\lfloor n/2 \rfloor} (-1)^k \binom{n}{k} \binom{2n-2k}{n} x^{n-2k} \quad n=0,1,\dots \quad (3.58.a)$$

Moreover, recurrence formulas associated with derivatives of Legendre polynomials are given by the relation

$$P'_{n+1}(x) - P'_{n-1}(x) = (2n+1)P_n(x), \quad n \geq 1. \quad (3.58.b)$$

Consider the following m^{th} order linear Fredholm Integro-differential equation with variable coefficients

$$\sum_{k=0}^m F_k(x) y^{(k)}(x) = g(x) + \lambda \int_{-1}^1 K(x,t) y(t) dt, \quad -1 \leq x, t \leq 1 \quad (3.59)$$

with conditions

$$\sum_{k=0}^{m-1} a_{jk} y^{(k)}(-1) + b_{jk} y^{(k)}(1) + c_{jk} y^{(k)}(0) = \mu_j, \quad j = 0, 1, 2, \dots, m-1 \quad (3.60)$$

where the constants a_{jk} , b_{jk} , c_{jk} , λ and μ_j are stable constants.

Our aim is to obtain a solution expressed in the form:

$$y(x) = \sum_{n=0}^N a_n P_n(x), \quad -1 \leq x \leq 1, \quad (3.61)$$

so that the Legendre coefficients to be determined are the a_n 's where $n=0,1,2,\dots,N$ and the functions $P_n(x)$ ($n=0,1,2,\dots,N$) are the Legendre polynomials defined by the formula (3.58.a). Here $F_k(x)$, $g(x)$ and $K(x,t)$ are functions defined in the interval $-1 \leq x, t \leq 1$.

Fundamental matrix relations

Equation (3.59) can be written as

$$D(x) = g(x) + \lambda I_f(x) \quad (3.62)$$

where

$$D(x) = \sum_{k=0}^m F_k(x) y^{(k)}(x)$$

and

$$I_f(x) = \int_{-1}^1 K(x,t) y(t) dt .$$

We convert the solution $y(x)$ and its derivative $y^{(k)}(x)$, the parts $D(x)$ and $I_f(x)$, and the mixed conditions in (3.60) to matrix form.

Matrix relations for $y(x)$ and $y^{(k)}(x)$

The function $y(x)$ defined by (3.61) can be expanded to the truncated Legendre series in the form

$$y(x) = \sum_{n=0}^N a_n P_n(x), \quad -1 \leq x \leq 1, \quad (3.63)$$

formula (3.63) and its derivative can be written in the matrix forms,

$$[y(x)] = P(x)A \quad \text{and} \quad [y^{(k)}(x)] = P^{(k)}(x)A, \quad (3.64)$$

where

$$P(x) = [P_0(x) \ P_1(x) \ \dots \ P_N(x)]$$

$$P^{(k)}(x) = [P^{(k)}(x) \quad P_1^{(k)}(x) \quad \dots \quad P_N^{(k)}(x)]$$

and
$$A = [a_0 \quad a_1 \quad \dots \quad a_N],$$

which A is the coefficient matrix.

On the other hand, by using the Legendre recursive formulas (3.58.a) and (3.58.b) and taking $n = 0, 1, \dots, N$ we can obtain the matrix equation

$$P^{(1)}(x) = P(x) \Pi^T, \quad (3.65)$$

where, for odd values of N

$$\begin{bmatrix} 0 & 0 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 1 & 0 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 0 & 3 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 1 & 0 & 5 & 0 & \dots & & 0 & & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & & & \vdots & & \vdots & \vdots \\ 0 & 3 & 0 & 7 & \dots & 2N-3 & & & 0 & 0 \\ 1 & 0 & 5 & 0 & \dots & & 0 & 2N-1 & & 0 \end{bmatrix}$$

for even values of N

$$\begin{bmatrix} 0 & 0 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 1 & 0 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 0 & 3 & 0 & 0 & \dots & & 0 & & 0 & 0 \\ 1 & 0 & 5 & 0 & \dots & & 0 & & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & & & \vdots & & \vdots & \vdots \\ 1 & 0 & 5 & 0 & \dots & 2N-3 & & & 0 & 0 \\ 0 & 3 & 0 & 7 & \dots & & 0 & 2N-1 & & 0 \end{bmatrix}$$

Also, it is clearly seen that the relation between the matrix $P(x)$ and its derivative $P^{(k)}(x)$, from (3.65), is

$$\begin{aligned}
P^{(1)}(x) &= P(x)\Pi^T \\
P^{(2)}(x) &= P^{(1)}(x)\Pi^T = P(x)(\Pi^T)^2 \\
P^{(k)}(x) &= P^{(k-1)}(x)(\Pi^T)^{k-1} = P(x)(\Pi^T)^k, k = 0,1,2,\dots
\end{aligned} \tag{3.66}$$

Consequently, by substituting the matrix relations (3.66) into Eq. (3.64), we obtain the matrix relations for $y(x)$ and $P^{(k)}(x)$ as

$$[y^{(k)}(x)] = P(x)(\Pi^T)^k A. \tag{3.67}$$

Matrix representations based on collocation points

The Legendre collocation points defined by

$$x_i = -1 + \frac{2}{N}i, \quad i = 0,1,\dots,N, \tag{3.68}$$

we substitute the collocation points (3.68) into Eq. (3.62) to obtain the system

$$D(x_i) = g(x_i) + \lambda I_f(x_i)$$

Represented in matrix equation as

$$D = G + \lambda I_f \tag{3.69}$$

where

$$D = \begin{bmatrix} D(x_0) \\ D(x_1) \\ \vdots \\ D(x_N) \end{bmatrix}, G = \begin{bmatrix} g(x_0) \\ g(x_1) \\ \vdots \\ g(x_N) \end{bmatrix}, I = \begin{bmatrix} I_f(x_0) \\ I_f(x_1) \\ \vdots \\ I_f(x_N) \end{bmatrix}$$

Matrix relation for the differential part $D(x)$

To reduce the $D(x)$ part to the matrix form by means of the collocation points, we first write the matrix D defined in eq.(3.69) as

$$D = \sum_{k=1}^m F_k Y^{(k)} \quad (3.70)$$

where

$$F_k = \begin{bmatrix} F_k(x_0) & 0 & \dots & 0 \\ 0 & F_k(x_1) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & F_k(x_N) \end{bmatrix},$$

$$Y^{(k)} = \begin{bmatrix} y^{(k)}(x_0) \\ y^{(k)}(x_1) \\ \vdots \\ y^{(k)}(x_N) \end{bmatrix}.$$

Using the collocation points x_i ($i = 0, 1, \dots, N$) in (3.67) we have the system of matrix equations

$$[y^{(k)}(x_i)] = P(x_i)(\Pi^T)^k A, \quad k = 0, 1, \dots, m$$

or

$$Y^{(k)} = \begin{bmatrix} y^{(k)}(x_0) \\ y^{(k)}(x_1) \\ \vdots \\ y^{(k)}(x_N) \end{bmatrix} = \begin{bmatrix} P(x_0) \\ P(x_1) \\ \vdots \\ P(x_N) \end{bmatrix} [(\Pi^T)^k A] = P(\Pi^T)^k A, \quad (3.71)$$

where

$$\mathbf{P} = \begin{bmatrix} P_0(x_0) & P_1(x_0) & \dots & P_N(x_0) \\ P_0(x_1) & P_1(x_1) & \dots & P_N(x_1) \\ \vdots & \vdots & & \vdots \\ P_0(x_N) & P_1(x_N) & \dots & P_N(x_N) \end{bmatrix}$$

Thus, from the matrix forms (3.70) and (3.71), we get the fundamental matrix relation for the differential part $D(x)$

$$D = \sum_{k=0}^m F_k P(\Pi^T)^k A. \quad (3.72)$$

Matrix relation for the Fredholm integral part $I_f(x)$

The kernel function $K(x, t)$ can be approximated by the truncated Legendre series

$$K(x, t) = \sum_{m=0}^N \sum_{n=0}^N k_{mn}^t P_m(x) P_n(t) \quad (3.73)$$

or

$$K(x, t) = \sum_{m=0}^N \sum_{n=0}^N k_{mn}^t x^m t^n \quad (3.74)$$

where

$$k_{mn}^t = \frac{1}{m!n!} \frac{\partial^{m+n} K(0,0)}{\partial x^m \partial t^n}; \quad m, n = 0, 1, \dots, N.$$

We convert (3.73) and (3.74) to matrix forms and then equalize

$$[K(x, t)] = P(x) K_f P^T(t) = X(x) K_f X^T(t) \quad (3.75)$$

where

$$P(x) = [P_0(x) \ P_1(x) \ \dots \ P_N(x)]$$

$$X(x) = [1 \ x \ \dots \ x^N]$$

$$K_l = [k_{mn}^l] \ , \ K_t = [k_{mn}^t]; \ m, n = 0, 1, \dots, N.$$

On the other hand, by using the Legendre recursive formula (3.58.a) for $n = 0, 1, \dots, N$ we can obtain the matrix equation

$$P^T(x) = DX^T(x) \quad \text{or} \quad P(x) = X(x)D^T \quad (3.76)$$

where for odd values of N and for even values of N see [42].

Therefore, substituting the matrix $P(x)$ in (3.76) into the relation (3.75)

and simplifying the result equation, we find the matrix relation

$$K_l = (D^{-1})^T K_t D^{-1} \quad (3.77)$$

which is the relation between the Legendre and Taylor coefficients of $K(x, t)$. Substituting the matrix forms (3.75) and (3.64) corresponding to the functions $K(x, t)$ and $y(t)$ into the Fredholm integral part $I_f(x)$, we obtain

$$[I_f(x)] = \int_{-1}^1 P(x) K_l P^T(t) P(t) A dt = P(x) K_l Q A, \quad (3.78)$$

where

$$Q = \int_{-1}^1 P^T(t) P(t) dt = [q_{mn}];$$

$$q_{mn} = \begin{cases} \frac{2N}{2m+1} & : m = n \\ 0 & : m \neq n \end{cases} ; m, n = 0, 1, \dots, N$$

and the matrix K_l is defined in (3.77).

Using the collocation points $x_i (i = 0, 1, \dots, N)$ defined in (3.68) in the relation (3.78) we obtain the system of the matrix equations

$$I_f(x_i) = P(x_i)_l K_l Q A; i = 0, 1, \dots, N$$

$$\text{or briefly} \quad I_f = PK_l Q A; \quad (3.79)$$

which is the fundamental matrix relation for $I_f(x)$.

Matrix relation for the mixed conditions

We can obtain the corresponding matrix form for the conditions (3.60), by means of the relation (3.66), as

$$\sum_{k=0}^{m-1} [a_{jk} P(-1) + b_{jk} P(1) + c_{jk} P(0)] (\Pi^T)^k A = \mu_j, j = 0, 1, \dots, m-1. \quad (3.80)$$

Method of solution

Now, we are ready to construct the fundamental matrix equation corresponding to (3.59). For this purpose, substituting the matrix relations (3.72) and (3.79) into Eq. (3.69) then simplifying it, we obtain the matrix equation

$$\sum_{k=0}^m \{F_k P(\Pi^T)^k - \lambda PK_l Q\} A = G, \quad (3.81)$$

which corresponds to a system of $(N + 1)$ algebraic equations for the $(N + 1)$ unknown Legendre coefficients a_0, a_1, \dots, a_N . Briefly, we can write Eq. (3.81) in the form

$$\mathbf{WA} = \mathbf{G} \quad \text{or} \quad [\mathbf{W} ; \mathbf{G}] \quad (3.82)$$

where

$$\mathbf{W} = [w_{pq}] = \sum_{k=0}^m F_k P(\Pi^T)^k - \lambda P K_l Q, \quad p, q = 0, 1, \dots, N$$

$$\mathbf{G} = [g(x_0) \quad g(x_1) \quad \dots \quad g(x^N)]^T.$$

On the other hand, the matrix form (3.80) for the conditions (3.60) can be written as

$$\mathbf{U}_j \mathbf{A} = [\mu_j] \quad \text{or} \quad [\mathbf{U}_j ; \mu_j], \quad j = 0, 1, \dots, m-1 \quad (3.83)$$

where

$$\mathbf{U}_j = \sum_{k=0}^{m-1} [a_{jk} P(-1) + b_{jk} P(1) + c_{jk} P(0)] (\Pi^T)^k$$

$$\equiv [u_{j0} \quad u_{j1} \quad \dots \quad u_{jN}].$$

To obtain the solution of Eq. (3.82) under the conditions (3.83), by replacing the rows matrices (3.83) by the last m rows of the matrix (3.81) we have the new augmented matrix

$$[\tilde{W}; \tilde{G}] = \begin{bmatrix} w_{00} & w_{01} & \cdots & w_{0N} & ; & g(x_0) \\ w_{10} & w_{11} & \cdots & w_{1N} & ; & g(x_1) \\ \vdots & \vdots & \ddots & \vdots & ; & \vdots \\ w_{N-m,0} & w_{N-m,1} & \cdots & w_{N-m,N} & ; & g(x_{N-m}) \\ u_{00} & u_{01} & \cdots & u_{0N} & ; & \mu_0 \\ u_{10} & u_{11} & \cdots & u_{1N} & ; & \mu_1 \\ \vdots & \vdots & \ddots & \vdots & ; & \vdots \\ u_{m-1,0} & u_{m-1,1} & \cdots & u_{m-1,N} & ; & \mu_{m-1} \end{bmatrix} \quad (3.84)$$

If $\text{rank } \tilde{W} = \text{rank}[\tilde{W}; \tilde{G}] = N + 1$, then we can write

$$A = (\tilde{W})^{-1} \tilde{G}.$$

Thus the coefficients $a_n (n = 0, 1, \dots, N)$ are uniquely determined by Eq.

(3.84).

3.4 Taylor Collocation Method

Consider the m th -order linear Fredholm integro-differential equation

$$\sum_{k=0}^m P_k(x) y^{(k)}(x) = f(x) + \lambda \int_a^b K(x, t) y(t) dt \quad (3.85)$$

where the unknown functions $P_k(x)$, $f(x)$, $K(x, t)$ are defined on $a \leq x, t \leq b$ and λ is a real parameter where $y(x)$ is the unknown function.

With conditions

$$\sum_{j=0}^{m-1} [a_{ij} y^{(j)}(a) + b_{ij} y^{(j)}(b) + c_{ij} y^{(j)}(c)] = \lambda_i, i = 0, 1, \dots, m-1 \quad (3.86)$$

where $a \leq c \leq b$, provided that the real coefficients a_{ij} , b_{ij} , c_{ij} and λ_i are appropriate constants.

We assume that the solution of (3.85) be the truncated Taylor series

$$y(x) = \sum_{n=0}^N \frac{y^{(n)}(c)}{n!} (x-c)^n; \quad a \leq x \leq b \quad (3.87)$$

where N is chosen any positive integer such that $N \geq m$. Then the solution (3.87) of Eq. (3.85) can be expressed in the matrix form

$$[y(x)] = X M_0 A$$

where

$$X = [1 \quad x-c \quad (x-c)^2 \quad \dots (x-c)^N]$$

$$A = [y^{(0)}(c) \quad y^{(1)}(c) \quad y^{(2)}(c) \quad \dots y^{(N)}(c)]$$

and

$$M_0 = \begin{bmatrix} \frac{1}{0!} & 0 & 0 & \dots & 0 \\ 0 & \frac{1}{1!} & 0 & \dots & 0 \\ 0 & 0 & \frac{1}{2!} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \frac{1}{N!} \end{bmatrix}.$$

To obtain a solution, we can use the following matrix method, which is a Taylor Collocation method. This method is based on computing the Taylor

coefficients by means of the Taylor collocation points are thereby finding the matrix A containing the unknown Taylor coefficients.

We substitute the Taylor collocation points

$$x_i = a + i \frac{b-a}{N}; \quad i = 0, 1, \dots, N; \quad x_0 = a, \quad x_N = b \quad (3.88)$$

into (3.85) to obtain

$$\sum_{k=0}^m P_k(x_i) y^{(k)}(x_i) = f(x_i) + \lambda I(x_i) \quad (3.89)$$

such that

$$I(x_i) = \int_a^b K(x, t) y(t) dt$$

then we can write the system (3.81) in the matrix form

$$P_0 Y^{(0)} + P_1 Y^{(1)} + \dots + P_m Y^{(m)} = \sum_{k=0}^m P_k Y^{(k)} = F + \lambda I \quad (3.90)$$

where

$$P_k = \begin{bmatrix} P_k(x_0) & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & P_k(x_N) \end{bmatrix}$$

$$\mathbf{F} = \begin{bmatrix} f(x_0) \\ f(x_1) \\ \vdots \\ f(x_N) \end{bmatrix}, \quad \mathbf{Y}^{(k)} = \begin{bmatrix} y^{(k)}(x_0) \\ y^{(k)}(x_1) \\ \vdots \\ y^{(k)}(x_N) \end{bmatrix}, \quad \mathbf{I} = \begin{bmatrix} I(x_0) \\ I(x_1) \\ \vdots \\ I(x_N) \end{bmatrix}.$$

Assume the k^{th} derivative of the function (3.87) with respect to x has the truncated Taylor series expansion defined by (3.87):

$$y^{(k)}(x_i) = \sum_{n=k}^N \frac{y^{(n)}(c)}{(n-k)!} (x_i - c)^{n-k}; \quad a \leq x \leq b$$

where $y^{(k)}(x)$ ($k = 0, \dots, N$) are Taylor coefficients; $y^{(0)}(x) = y(x)$.

Then substituting the Taylor collocation points, we obtain

$$[y^{(k)}(x_i)] = \mathbf{X}_{x_i} \mathbf{M}_k \mathbf{A}, \quad k = 0, 1, \dots, N \quad (3.91)$$

or the matrix equation

$$\mathbf{y}^{(k)} = \mathbf{C} \mathbf{M}_k \mathbf{A} \quad (3.92)$$

where

$$\mathbf{C} = \begin{bmatrix} \mathbf{X}_{x_0} \\ \mathbf{X}_{x_1} \\ \vdots \\ \mathbf{X}_{x_N} \end{bmatrix} = \begin{bmatrix} (x_0 - c)^0 & (x_0 - c)^1 & \dots & (x_0 - c)^N \\ (x_1 - c)^0 & (x_1 - c)^1 & \dots & (x_1 - c)^N \\ \vdots & \vdots & \ddots & \vdots \\ (x_N - c)^0 & (x_N - c)^1 & \dots & (x_N - c)^N \end{bmatrix}$$

$$\mathbf{M}_k = \begin{bmatrix} 0 & 0 & \dots & \frac{1}{0!} & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & \frac{1}{1!} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 0 & 0 & \frac{1}{(N-k)!} \\ 0 & 0 & \dots & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & 0 & \dots & 0 \end{bmatrix}.$$

Then Eq. (3.88) can be written as

$$\left(\sum_{k=0}^m \mathbf{P}_k \mathbf{C} \mathbf{M}_k \right) \mathbf{A} = \mathbf{F} + \lambda \mathbf{I} \quad (3.93)$$

The kernel $K(x, t)$ is expanded to construct Taylor series

$$K(x, t) = \sum_{n=0}^N \sum_{m=0}^N k_{nm} (x-c)^n (t-c)^m$$

$$k_{nm} = \frac{1}{n!m!} \left. \frac{\partial^{n+m} K(0,0)}{\partial x^n \partial t^m} \right|_{(x=c, t=c)}.$$

The matrix representation of $K(x, t)$ defined by

$$[K(x, t)] = \mathbf{X} \mathbf{K} \mathbf{T}^T \quad (3.94)$$

where

$$\mathbf{X} = [1 \quad x-c \quad (x-c)^2 \quad \dots (x-c)^N]$$

$$\mathbf{T} = [1 \quad t-c \quad (t-c)^2 \quad \dots (t-c)^N]$$

$$\mathbf{K} = \begin{bmatrix} k_{00} & k_{01} & \dots & k_{0N} \\ k_{10} & k_{11} & \dots & k_{1N} \\ \vdots & \vdots & \ddots & \vdots \\ k_{N0} & k_{N1} & \dots & k_{NN} \end{bmatrix}$$

In addition, the matrix representation of $y(x)$ and $y(t)$ are

$$[y(x)] = \mathbf{X} \mathbf{M}_0 \mathbf{A}, \quad [y(t)] = \mathbf{T} \mathbf{M}_0 \mathbf{A}. \quad (3.95)$$

Setting (3.94) and (3.95) into $I(x_i)$ defined in (3.89), we have

$$[I(x)] = \int_a^b \{ \mathbf{X} \mathbf{K} \mathbf{T}^T \mathbf{T} \mathbf{M}_0 \mathbf{A} \} dt = \mathbf{X} \mathbf{K} \mathbf{H} \mathbf{M}_0 \mathbf{A} \quad (3.96)$$

$$\mathbf{H} = [h_{nm}] = \int_a^b \mathbf{T}^T \mathbf{T} dt, \quad h_{nm} = \left. \frac{(b-c)^{n+m+1} - (a-c)^{n+m+1}}{n+m+1} \right|_{n,m=0,1,\dots,N}$$

from (3.96) we get the matrix I in the form

$$\mathbf{I} = \mathbf{C} \mathbf{K} \mathbf{H} \mathbf{M}_0 \mathbf{A}. \quad (3.97)$$

Finally, substituting (3.97) in (3.93), we get the matrix equation

$$\left(\sum_{k=0}^m \mathbf{P}_k \mathbf{C} \mathbf{M}_k - \lambda \mathbf{C} \mathbf{K} \mathbf{H} \mathbf{M}_0 \right) \mathbf{A} = \mathbf{F} \quad (3.98)$$

which is the fundamental relation for solving of Fredholm integro-differential equation defined in $a \leq x \leq b$.

Equation (3.98) can be written as

$$\mathbf{W} \mathbf{A} = \mathbf{F} \quad (3.99)$$

which corresponds to a system of $(N + 1)$ algebraic equations with the unknown Taylor coefficients

$$W = [w_{ij}] = \sum_{k=0}^m P_k C M_k - \lambda C K H M_0, \quad i, j = 0, 1, \dots, N.$$

Also, for the conditions in (3.86). We assume

$$\begin{aligned} [y^{(j)}(a)] &= P M_j A \\ [y^{(j)}(b)] &= Q M_j A \\ [y^{(j)}(c)] &= R M_j A \end{aligned} \quad (3.100)$$

where

$$P = [1 \quad (a-c) \quad (a-c)^2 \quad \dots (a-c)^N]$$

$$Q = [1 \quad (b-c) \quad (b-c)^2 \quad \dots (b-c)^N]$$

$$R = [1 \quad 0 \quad 0 \quad \dots 0].$$

Setting (3.100) in (3.86), we get

$$\sum_{j=0}^{m-1} \{a_{ij} P + b_{ij} Q + c_{ij} R\} M_j A = [\lambda_i]$$

Let

$$U_i = \sum_{j=0}^{m-1} \{a_{ij} P + b_{ij} Q + c_{ij} R\} M_j = [u_{i0} \quad u_{i1} \quad \dots u_{iN}], \quad i = 0, 1, \dots, m-1. \quad (3.101)$$

Thus, the conditions (3.86) becomes

$$U_i A = [\lambda_i] \quad (3.102)$$

also the augmented matrices representing them are

$$[U_i; \lambda_i] = [u_{i0} \ u_{i1} \ \dots \ u_{iN}; \lambda_i]. \quad (3.103)$$

Finally, the augmented matrix $[W; F]$ is defined by the matrices:

$$W = \begin{bmatrix} w_{00} & w_{01} & \dots & w_{0N} \\ w_{10} & w_{11} & \dots & w_{1N} \\ \vdots & \vdots & \ddots & \vdots \\ w_{N-m,0} & w_{N-m,1} & \dots & w_{N-m,N} \\ u_{00} & u_{01} & \dots & u_{0N} \\ \vdots & \vdots & \ddots & \vdots \\ u_{m-1,0} & u_{m-1,1} & \dots & u_{m-1,N} \end{bmatrix}$$

$$F = \begin{bmatrix} f(x_0) \\ f(x_1) \\ \vdots \\ f(x_{N-m}) \\ \lambda_0 \\ \vdots \\ \lambda_{m-1} \end{bmatrix}.$$

If, $|W| \neq 0$ we can write

$$A = W^{-1} F$$

Matrix A is uniquely determined. Then (3.85) has a unique solution in the form (3.87).

Chapter Four

Numerical Examples and Results

To test the efficiency and accuracy of the numerical methods represented in chapter three we will consider the following numerical examples:

Example 4.1

Consider the Fredholm integro-differential equation of the second kind

$$u'(x) = 3e^{3x} - \frac{1}{3}(2e^3 + 1)x + \int_0^1 3xt u(t) dt \quad \text{with } u(0) = 1. \quad (4.1)$$

Equation (4.1) has the exact solution $u(x) = e^{3x}$.

4.1 The numerical realization of equation (4.1) using the

B-spline scaling functions and wavelets on [0,1]

The following algorithm implements the **B-spline scaling functions and wavelets on [0,1]** using the **Maple** software

Algorithm 4.1 [25]

1. Input the fixed positive integer M
2. Input the values of the matrix Q_{M+1}
3. Input the values of the matrix R
4. Input the values of the matrix $\alpha(j)$
5. Input the values of the matrix $\beta(j)$

6. Calculate the matrix G , by using (3.24) and (3.38).

7. Calculate the matrix D_1

8. For i from 2 to $2^{M+2} - 2$, $X(1) = 0$, $X(2^{M+1} + 1) = 1$

$$X(i) = (2i - 1) / (2^{M+2} - 2)$$

9. Defined $\chi_{i,j}(x)$, $\chi_{i,-1}(x)$, $\chi_{i,2^i-1}(x)$

10. Defined $\mu_{i,j}(x)$, $\mu_{i,-1}(x)$, $\mu_{i,2^i-1}(x)$

11. Input $h = \frac{(b-a)}{M}$, $t(i) = i * h$

12. For i from 1 to 5, for j from 1 to $2^{M+1} + 1$, for k from 2 to M , for I from 1 to 2^K , define $mut(i, j) = (mu(2, i - 2, t(j - 1)))$, and $mut_1(i, j) = (mu(k, I - 2, (j - 1)))$ to calculate $mutt = \langle mut, mut_1 \rangle$

13. Define the kernel

14. For i from 1 to 5, for j from 1 to $2^{M+1} + 1$, for k from 2 to M , for I from 1 to 2^K , define $muX(i, j) = (chi(2, i - 2, X(j)))$, and $muX_1(i, j) = (mu(K, I - 2, X(j)))$

15. Calculate $muXX = \langle muX, muX_1 \rangle$

16. Calculate $zz = z - \text{integration}$

17. $F(1,1) = 1$, and calculate $F(1,i)$

18. Calculate $C = F * \text{Matrix Inverse } zz$

19. The solution is $U = C * muXX$

20. Input the exact solution

$$u = \text{Transpose}(U)$$

$$u_1 = \text{convert}(u, \text{vector})$$

$$r := \text{plot}(X, u_1)$$

$$r_2 := \text{plot}(u_2(x))$$

21. Display (r, r_2)

22. Absolute error = $|u - u_M|$

Table 4.1 shows the exact and numerical solutions when applying algorithm 4.1 on equation (4.1), and showing the resulting error of using the numerical solution.

Table 4.1: The exact and numerical solutions of applying Algorithm 4.1 on equation (4.1).

x	Exact solution $u(x) = \exp(3x)$	Numerical solution u_M	Absolute error $= u - u_M $
0.1	1.349858808	1.36616469332652	0.0163058853265190
0.2	1.822118800	1.85217464520521	0.0300558452052084
0.3	2.459603111	2.48804090541187	0.0284377944118681
0.4	3.320116923	3.33272256083022	0.012605637830226
0.5	4.4816890701	4.51948395416716	0.0377948841671625
0.6	6.049647464	6.06159218771469	0.0119447237146941
0.7	8.166169913	8.19415390871779	0.0279839957177916
0.8	11.02317638	11.0673029532516	0.0441265732516403
0.9	14.87993172	14.8583456549649	0.0213860650350881
1.0	20.08553692	19.7454628997592	0.340074020240781

These results show the accuracy of the B-spline scaling functions and wavelets to solve equation (4.1) with the max error = 0.34007402024078.

Figure 4.1 compares the exact solution $u(x) = e^{3x}$ with the approximate solution with $M = 4$.

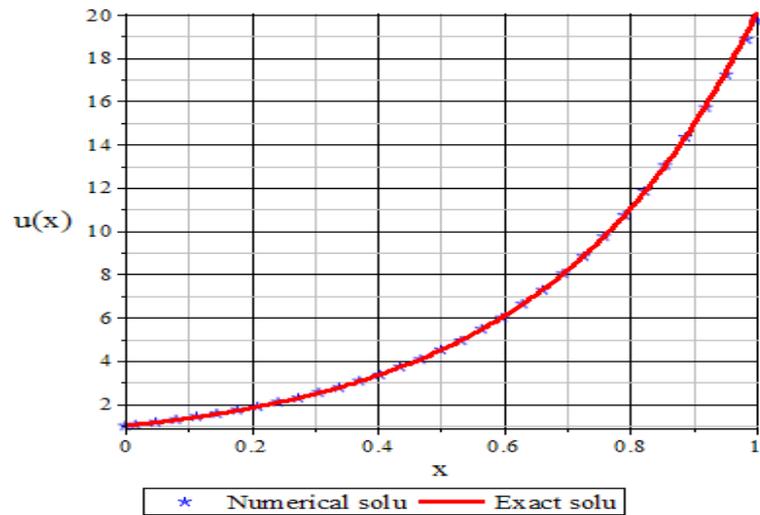


Figure 4.1: The exact and numerical solutions of applying algorithm 4.1 for equation (4.1).

Figure 4.2 shows the absolute error resulting of applying algorithm 4.1 on equation (4.1).

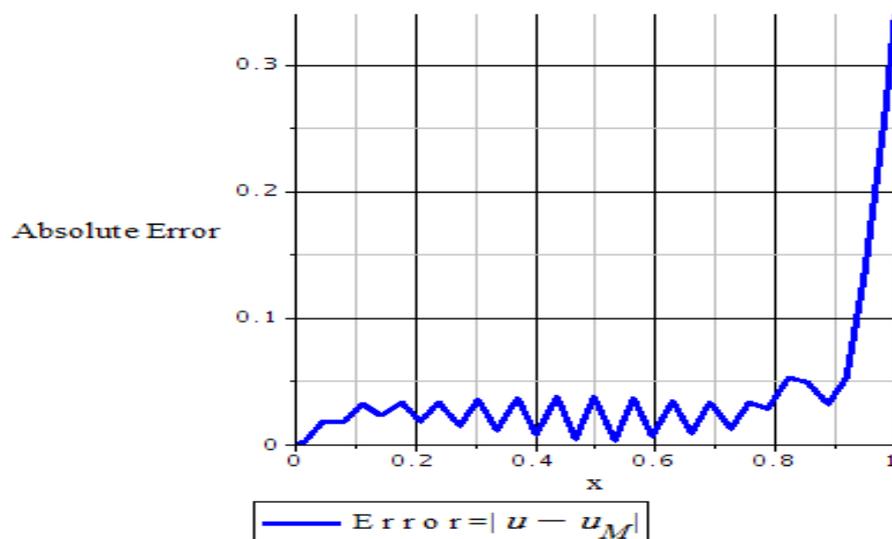


Figure 4.2: The error resulting of applying algorithm 4.1 on equation (4.1).

4.2 The numerical realization of equation (4.1) using the Homotopy perturbation method

The following algorithm implements the **Homotopy perturbation method** using the **Maple** software

Algorithm 4.2

- (1) Input $a, \lambda, n, f(x), G(x, y)$
- (2) Calculate $wm1[0] = \text{int}((f, x))$
- (3) Calculate $wm[0] := y \rightarrow \text{sub}(x = y, wm1[0])$
- (4) Calculate $c = a - \text{eval}(wm0)$
- (5) Define $ws[0] = wm1[0] + c$
- (6) Calculate $w[0] = \text{subs}(x = y, ws[0])$
- (7) For $i = 1$ to n , calculate
 - $wm[i] = (\text{int}(G * w[i - 1], y = 0..1))$
 - $ws[i] = (\text{int}(wm[i], x))$
 - $w[i] = \text{subs}(x = y, ws[i])$
- (9) Define $u_n = \text{combine}(\text{add}(ws[k], k = 0..1))$,
- (10) Define
 - $u := x \rightarrow \exp(x)$
 - $r := \text{plot}(u_n(x))$
 - $r_2 := \text{plot}(u(x))$
- (11) Absolute Error = $|u - u_n|$

For more details see [21, 43].

Table (4.2) shows the exact and numerical results when $n = 4$, and showing the error resulting of using the numerical solution.

Table (4.2): The exact and numerical solution of applying Algorithm 4.2 on equation (4.1).

x	Exact solution $u(x) = e^{3x}$	Numerical solution $u_n(x)$	Absolute Error $= u - u_n $
0.1	1.349858808	1.348501851	0.001356957
0.2	1.822118800	1.816690973	0.005427827
0.3	2.459603111	2.447390500	0.012212611
0.4	3.320116923	3.298405614	0.021711309
0.5	4.4816890704	4.447765150	0.033923920
0.6	6.049647463	6.000797020	0.048850444
0.7	8.166169918	8.099679031	0.066490882
0.8	11.023176388	10.936331144	0.08684524
0.9	14.879731722	14.679818222	0.10991350
1	20.085536922	19.949841244	0.13569568

These results show the accuracy of the Homotopy perturbation method to solve equation (4.1) with the max error = 0.13569568.

Figure 4.3: shows both exact and numerical solutions with $n = 4$.

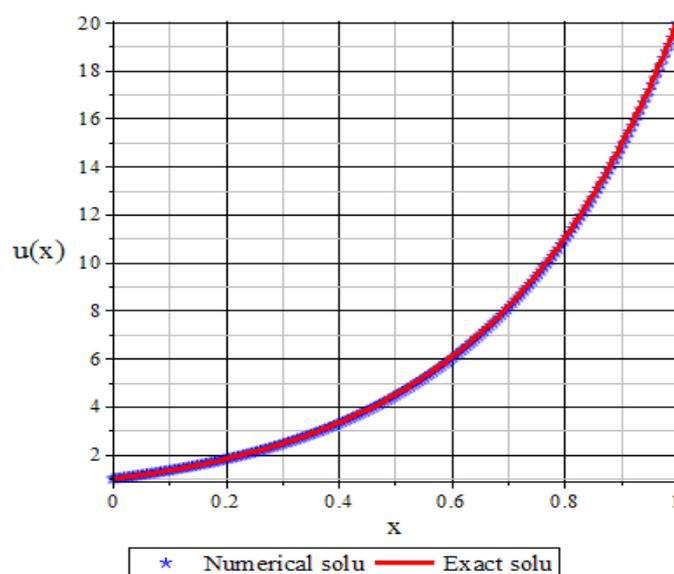


Figure 4.3: The exact and numerical solutions of applying algorithm 4.2

for equation (4.1)

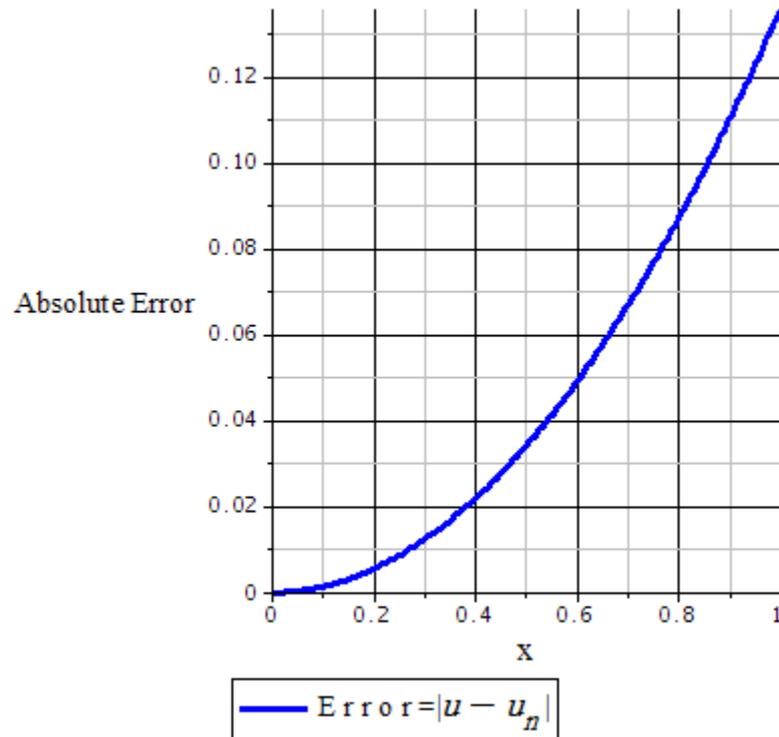


Figure 4.4: The error resulting of applying algorithm 4.2 on equation (4.1).

Example 4.2

Consider the Fredholm integro-differential equation

$$u'(x) = -2\pi \sin(2\pi x) - \pi \sin(4\pi x) + 2\pi \int_0^1 \sin(4\pi x + 2\pi t) u(t) dt, \quad (4.2)$$

$$u(0) = 1.$$

Equation (4.2) has the exact solution $u(x) = \cos(2\pi x)$.

4.3 The numerical realization of equation (4.2) using the B-spline scaling functions and wavelets on [0,1]

Table (4.3) shows the exact and numerical results when $M = 8$, and showing the error resulting of using the numerical solution.

Table (4.3): The exact and numerical solution of applying Algorithm 4.1 on equation (4.2).

x	Exact solution $u(x) = \cos(2\pi x)$	Approximate solution u_M	Aboslute Error $ u - u_M $
0.1	0.8090169943	0.716839678443304	0.0921773158566965
0.2	0.3090169938	0.198368704577070	0.11064828932222930
0.3	-0.3090169938	-0.414317811956477	0.105300818156477
0.4	-0.8090169943	-0.909564138036594	0.100547143736594
0.5	-1	-1.09848238174981	0.0984823817498084
0.6	0.8090169943-	-0.910024012119846	0.101007017819846
0.7	0.3090169938-	-0.414601813875286	0.105584820075286
0.8	0.3090169938	0.198637238538349	0.1110379755261651
0.9	0.8090169943	0.717312752695359	0.0917042416046411
1	1	1.00027574181029	0.000275741810288777

These results show the accuracy of the B-spline scaling functions and wavelets to solve equation (4.2) with the max error =0.11064828932222930.

Figure 4.5 shows both the exact and the numerical solutions with $M = 8$.

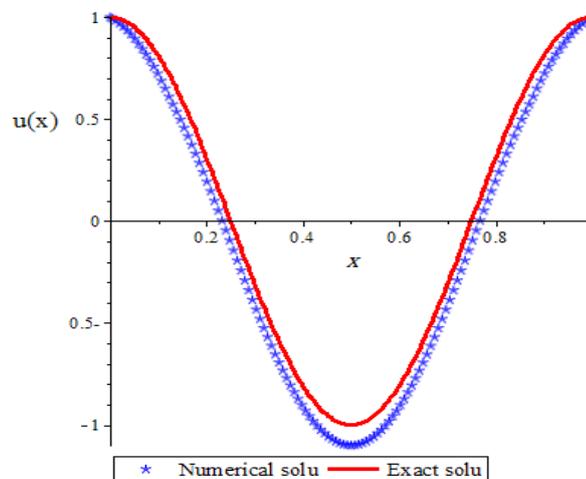


Figure 4.5: The exact and numerical solutions of applying algorithm 4.1 on equation (4.2).

Figure 4.6 shows the absolute error resulting of applying algorithm 4.1 on equation (4.2).

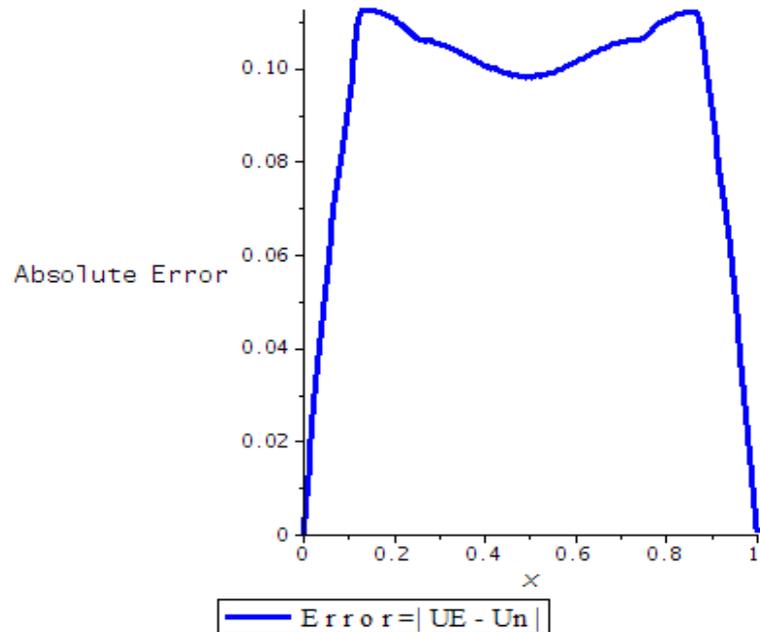


Figure 4.6 : The error resulting of applying algorithm 4.1 on equation (4.2).

4.4 The numerical realization of equation 4.2 using the Homotopy perturbation method

Table (4.4) shows the exact and numerical solutions when applying algorithm 4.2 on equation (4.2) when $n = 8$, and showing the error resulting of using the numerical solution.

Table (4.4): The exact and numerical solution of applying Algorithm 4.2 on equation (4.2).

x	Exact solution $u(x) = \cos(2\pi x)$	Approximate solution $u_n(x)$	Error $= u - u_n $
0	1	1	0
0.1	0.8090169943	0.8090169943	0
0.2	0.3090169938	0.3090169938	0
0.3	0.3090169938-	0.3090169942-	4×10^{-10}
0.4	0.8090169943-	0.8090169945-	2×10^{-10}
0.5	-1	-1	0
0.6	0.8090169943-	0.8090169940-	3×10^{-10}
0.7	0.3090169938-	0.3090169934-	4×10^{-10}
0.8	0.3090169938	0.3090169946	8×10^{-10}
0.9	0.8090169943	0.8090169947	4×10^{-10}
1	1	1	0

These results show the accuracy of the Homotopy perturbation method to solve equation (4.2) since the max error = 8×10^{-10} .

Figure 4.7: shows both exact and the numerical solutions with $n = 8$

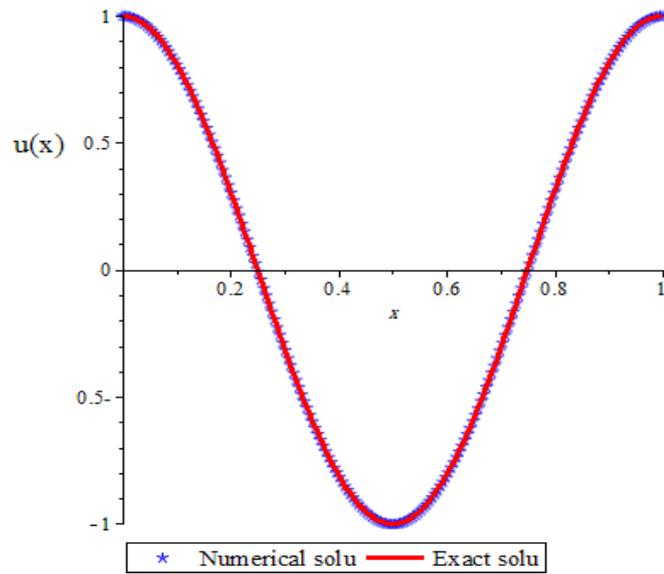


Figure 4.7: The exact and numerical solutions of applying algorithm 4.2 on equation (4.2).

Figure 4.8 shows the absolute error resulting of applying algorithm 4.2 on equation (4.2)

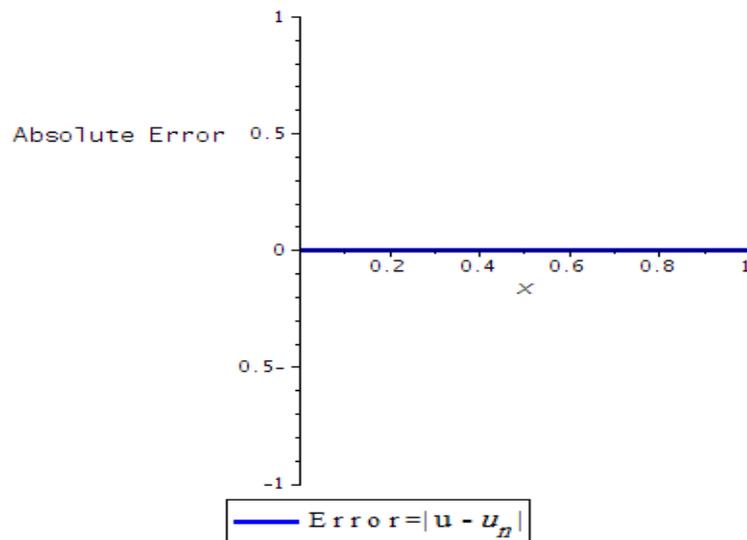


Figure 4.8 : The error resulting of applying algorithm 4.2 on equation (4.2).

Example 4.3

Consider the Fredholm integro-differential equation

$$y'' + xy' - xy = e^x - 2\sin(x) + \int_{-1}^1 \sin(x)e^{-t}y(t)dt, y(0) = 1, y'(0) = 1 \quad (4.3)$$

Equation (4.3) has the exact solution $y(x) = \exp(x)$.

We will find an approximate solution to equation (4.3) by the Legendre polynomial method, the Taylor collocation method.

4.5 The numerical realization of equation 4.3 using the Legendre polynomial method

The following algorithm implements the **Legendre polynomial method** using the **Matlab** software

Algorithm 4.3 [42]

1. Input $N, a, b, m, P_0(x), P_1(x), P_2(x), f(x), G(x, y)$

2. Let $x_i = \text{zeros}(1, N + 1)$

for $i = 0:N$

$$x_i(1, i + 1) = a + i * ((b - a) / N)$$

end

3. Let $F = \text{zeros}(N + 1, N + 1, N)$

for $k = 1:\text{length}(f)$

for $i = 0:N$

$$F(i+1, i+1, k) = \text{subs}(f(k), \{x\}, \{x_i(i+1)\})$$

end

end

4. $Pk = \text{zeros}(N)$

for $j = 0:N$

for $i = 0:N$

$$P = \text{Legendrep}(i, 'X')$$

$$X = x_i(j+1)$$

$$Pk(j+1, i+1) = \text{evalf}(P)$$

end

end

5. Define $PI = \text{LegendrePi}(N+1)$

6. Define $K_t = \text{LegendreK}_t(N, K(x, t))$

7. Calculate $Q = \text{diag}(2 / (2 * n + 1))$

8. Calculate $G = \text{subs}(g(x), x, x_i)$

9. Let $WD = \text{zeros}(N+1, N+1, N-1)$

for $k = 0:N-1$

$$WD(:, :, k + 1) = F(:, :, k + 1) * Pk * ((PI')^k)$$

end

10. Let $WX = \text{zeros}(N + 1)$

for $k = 0:N - 1$

$$WD(:, :) = WX(:, :) + WD(:, :, k + 1)$$

end

$$W = WX - \lambda * Pk * K_1 * Q$$

11. Let $U_0 = \text{zeros}(1, N)$

for $i = 0:N$

$$P = \text{Legendre}(i, 'X')$$

$$X = 0$$

$$U_0(1, i + 1) = \text{evalf}(P)$$

end

12. Let $U_1 = \text{zeros}(1, N)$

$$U(1, 1) = 0$$

for $i = 0:N$

$$P = \text{Legendrep}(i, 'X')$$

$$P = \text{evalf}(P)$$

$$P = \text{diff}(P)$$

$$U_1(1, i + 1) = \text{subs}(P, x, 0)$$

end

$$13. \text{ Put } WU = [W(1:N-1, :); U_0, U_1]$$

$$14. \quad GU = [G(1:N-1), 1, 1]$$

$$15. \quad A = WU / GU$$

$$16. \quad Y = '0'$$

for $i = 0:N$

$$aux = ['num2str(A(i+1))$$

$$\text{Legendrep}(i, 'x')']$$

$$Y = [Y '+' aux]$$

end

$$Y = \text{evalf}(Y)$$

Table 4.5 shows the exact and numerical solutions when applying algorithm 4.3 on equation (4.3), and showing the error resulting of using the numerical solution.

Table 4.5: The exact and numerical solution of applying Algorithm 4.3 on equation (4.3).

x_i	Exact solution $y(x) = \exp(x)$	Approximate solution $y_N(x)$	Absolute Error = $ y - y_N $
-1	0.36787944	0.36787945	1×10^{-8}
-0.8	0.44932896	0.44932895	1×10^{-8}
-0.6	0.54881163	0.54881163	0
-0.4	0.67032004	0.67032005	1×10^{-8}
-0.2	0.81873075	0.81873075	0
0	1	1	0
0.2	1.22140275	1.22140276	1×10^{-8}
0.4	1.49182469	1.49182470	0
0.6	1.82211880	1.82211882	2×10^{-8}
0.8	2.22554092	2.22554046	4.6×10^{-7}
1	2.71828182	2.71827657	5.25×10^{-6}

These results show the accuracy of the Legendre polynomial method to solve equation (4.3) with the max error = 5.25×10^{-6} .

Figure 4.9 compares the exact solution $y(x) = \exp(x)$ with the approximate solution

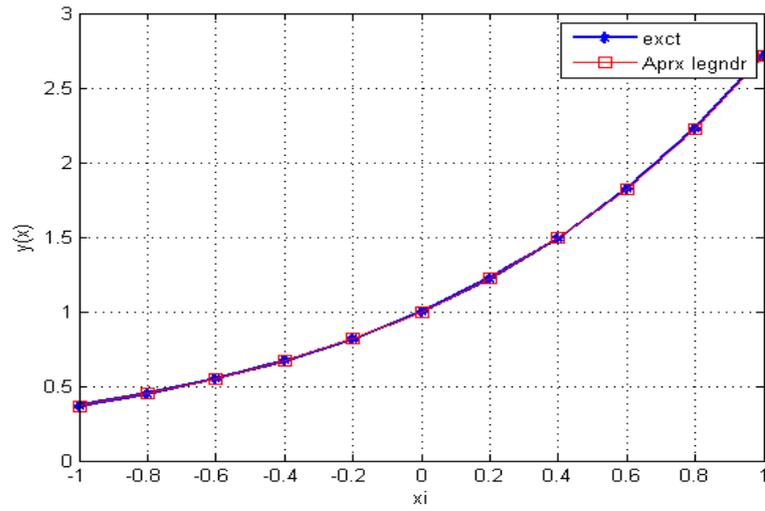


Figure 4.9: The exact and numerical solutions of applying algorithm 4.3 on equation (4.3).

Figure 4.10 shows the absolute error resulting of applying algorithm 4.3 on equation (4.3)

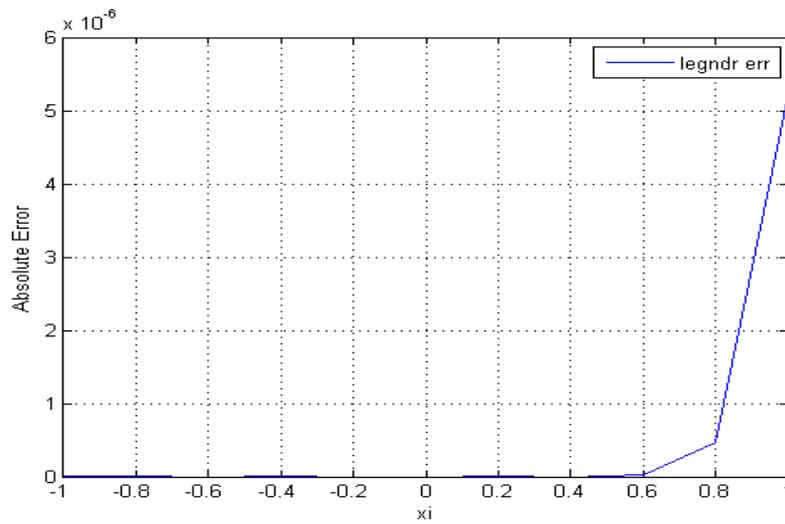


Figure 4.10: The error resulting of applying algorithm 4.3 on equation (4.3).

4.6 The numerical realization of equation 4.3 using the Taylor collocation method

The following algorithm implements the **Taylor collocation method** using the **Matlab** software

Algorithm 4.4

1. Input $a, b, c, K(x, t), f(x), P_0, P_1, P_2$

2. Let $x_i = \text{zeros}(1, N + 1)$

for $i = 0:N$

$$x_i(1, i + 1) = a + i * ((b - a) / N)$$

end

3. Let $H = \text{zeros}(N + 1)$

for $n = 0:N$

for $m = 0:N$

$$\text{expnt} = n + m + 1$$

$$H(n + 1, m + 1) = ((b - c) ^ \text{expnt} - (a - c) ^ \text{expnt}) / \text{expnt}$$

end

end

4. Let $C = \text{zeros}(N + 1)$

for $n = 0:N$

for $m = 0:N$

$$C(n + 1, m + 1) = (x_i(1, n + 10))^m$$

end

end

5. Let $F = \text{zeros}(N + 1, 1)$

for $i = 0:N$

$$F(i + 1, 1) = \text{subs}(FX, \{x\}, \{x_i(i + 1)\})$$

end

6. Let $K_i = \text{zeros}(N + 1)$

for $n = 0:N$

for $m = 0:N$

for $td = 0:m$

if $td > 0$

$$PK = \text{diff}(K(x, t), t)$$

$$K(x, t) = Pk$$

end

end

for $xd = 0:n$

if $xd > 0$

$$PK = \text{diff}(K(x,t), x)$$

$$K(x,t) = Pk$$

end

end

7. Calculate $K_l(n+1, m+1) = (1 / (\text{factorial}(n) * \text{factorial}(m)))$

$$* \text{subs}(K(x,t), \{x,t\}, \{0,0\})$$

8. Input the values of the matrix M

9. Input $P_0 = -x$, $P_1 = x$, $P_2 = 1$, $P_S = [1, x, -x]$

$$P = \text{zeros}(N+1, N+1, K)$$

for $k = 0:K$

for $n = 0:N$

for $m = 0:N$

if $n = m$

$$P(n+1, m+1, (k=1)) = \text{subs}(P_S(k+1), \{x\}, \{x_i(m+1)\})$$

end

end

end

end

10. Let $YS = \text{zeros}(N+1, 1)$

$$YS(1) = 1$$

$$YS(2) = 1$$

11. $FU = [F; YS]$

12. Let $U = \text{zeros}(N+1)$

for $n = 0:N$

for $m = 0:N$

if $n = m$ & $(YS(m+1) = 1)$

end

end

end

13. Let $WD = \text{zeros}(N+1, N+1, K)$

for $k = 0:K$

$$WD(:, :, k + 1) = P(:, :, k + 1) * C * M(:, :, k + 1)$$

end

14. Let $WX = \text{zeros}(N + 1)$

for $k = 0:K$

$$WD(:, :) = WX(:, :) + WD(:, :, k + 1)$$

end

15. Calculate $W = WX - \lambda * C * K_l * H * M(:, :, 1)$

$$WU = [W ; U]$$

16. $A = WU / FU$

17. Put $Y = 0$

for $n = 0:N$

$$Y = Y + (A(n + 1) * \text{factorial}(n)) * (x - c)^n$$

end

Table 4.6 shows the exact and numerical results , and showing the error resulting of using the numerical solution.

Table 4.6: The exact and numerical solution of applying Algorithm 4.4 on equation (4.3).

x_i	Exact solution $y(x) = \exp(x)$	Approximate solution $y_N(x)$	Error = $ y - y_N $
-1	0.367879	0.368019	1.4×10^{-4}
-0.8	0.449328	0.449403	7.5×10^{-5}
-0.6	0.548811	0.548854	4.3×10^{-5}
-0.4	0.670320	0.670347	2.7×10^{-5}
-0.2	0.818730	0.818741	1.1×10^{-5}
0	1	1	0
0.2	1.221402	1.221415	1.3×10^{-5}
0.4	1.491824	1.491827	3.0×10^{-6}
0.6	1.822118	1.821847	2.7×10^{-4}
0.8	2.225540	2.224079	1.5×10^{-3}
1	2.718281	2.713341	4.9×10^{-3}

These results show the accuracy of the Taylor collocation method to solve equation (4.3) with a maximum error = 4.9×10^{-3} .

Figure 4.11 compares the exact solution $y(x) = \exp(x)$ with the approximate solution.

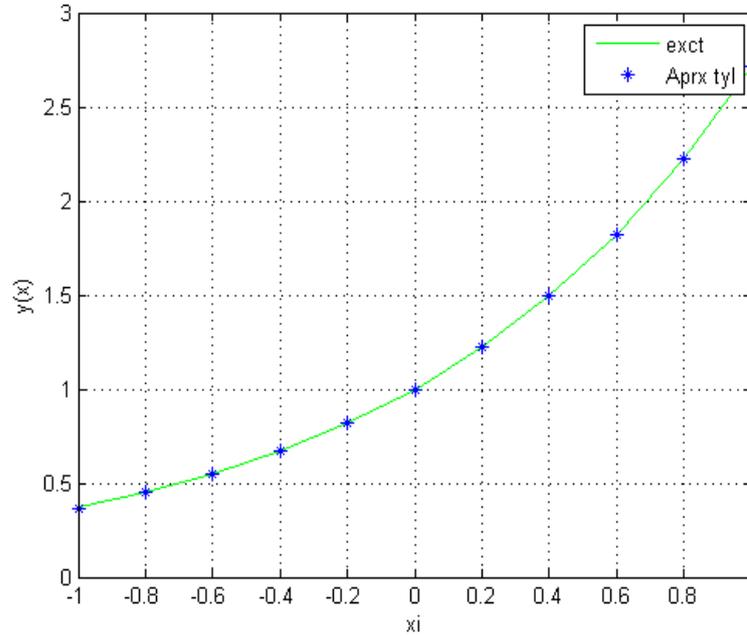


Figure 4.11: The exact and numerical solutions of applying algorithm 4.4 on equation (4.3).

Figure 4.12 shows the absolute error resulting of applying algorithm 4.4 on equation (4.3)

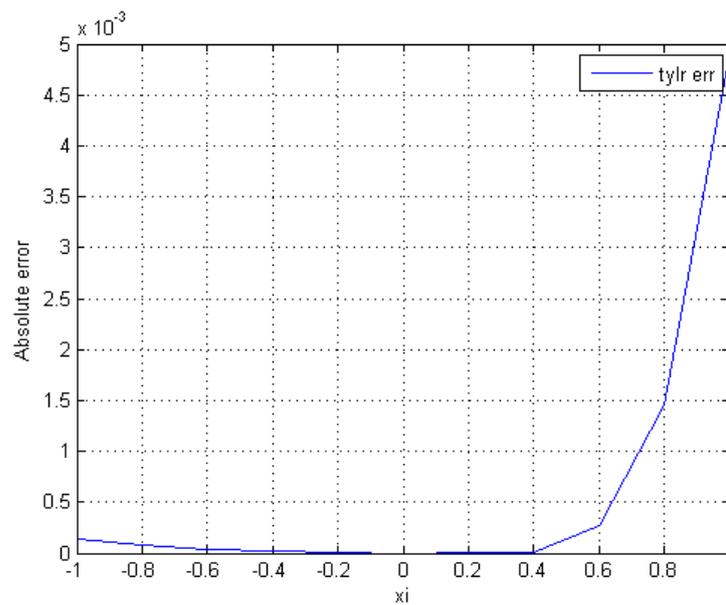


Figure 4.12: The error resulting of applying algorithm 4.4 on equation (4.3).

Conclusions

Analytical and numerical methods have been used to solve linear Fredholm integro-differential equation of the second kind. The numerical methods are implemented in a form of algorithms to solve some numerical examples using Maple and Matlab software.

The numerical results show the following observations:

- (1) Numerical results for examples 4.1 and 4.2 show clearly that the Homotopy Perturbation Method (HPM) is very efficient in comparison with the B-spline scaling functions and wavelets method. This is because the HPM is very well known for its fast convergence and consequently requires less CPU time and therefore less error. Moreover, the HPM introduces less complexity and can easily be implemented.
- (2) Example 4.2 has been solved numerically by both the Legendre polynomial method and the Taylor collocation method. In fact one of the most important features of the Legendre polynomial method is that the error introduced is considerably small in comparison with the Taylor collocation method.

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Appendix

Maple Code for B-Spline Scaling Functions and Wavelets for

Example 4.1

```

> restart:

> with(plots):with(LinearAlgebra):

> M:=4:

> 2^(M+1)+1;

> Q:=Matrix(2^(M+1)+1):R:=Matrix(2^(M+1)+1):

> for i from 1 to (2^(M+1)+1) do
> for j from 1 to (2^(M+1)+1) do
> if (i=j+1) then Q(i,j):= 1/24; R(i,j):= 1/2 elif
(j=i+1)then Q(i,j):= 1/24;R(i,j):= -1/2 elif (i=j) then
Q(i,j):= 1/6 end if
> od;

> od:

Q(1,1):=1/12:Q(2^(M+1)+1,2^(M+1)+1):=1/12:R(1,1):=-
1/2:R(2^(M+1)+1,2^(M+1)+1):=1/2:

> Qm:= 1/2^(M-1)*Q:

> for j from 2 to M do

> alpha(j):= Matrix(2^j+1,2^(j+1)+1):

> beta(j):=Matrix(2^j,2^(j+1)+1):

> od:

> for k from 2 to M do

> for i from 1 to 2^k+1 do

> for j from 1 to 2^(k+1)+1 do

> if (j=2*i-1) then alpha(k)(i,j):= 1 elif (j=2*(i-1))then
alpha(k)(i,j):= 1/2 elif (j=2*i) then alpha(k)(i,j):= 1/2
end if

> od:

> od:

```

```

> od:

> od:

> for k from 2 to M do
> for i from 2 to 2^k-1 do
> for j from 1 to 2^(k+1)+1 do
> if (j=2*i-1) then beta(k)(i,j):= -1/2 elif (j=2*(i-
1))then beta(k)(i,j):= 1/12 elif (j=2*i) then
beta(k)(i,j):= 5/6 elif (j=2*i+1) then beta(k)(i,j):= -1/2
elif (j=2*(i+1)) then beta(k)(i,j):=1/12 end if
> od:

> od:

> beta(k)(1,1):= -1:beta(k)(1,2):= 11/12:beta(k)(1,3):= -
1/2:beta(k)(1,4):= 1/12:beta(k)(2^k,2^(k+1)+1):= -
1:beta(k)(2^k,2^(k+1)):= 11/12:beta(k)(2^k,2^(k+1)-1):= -
1/2:beta(k)(2^k,2^(k+1)-2):= 1/12:

> od:

> for i from 2 to M do
> q(i):=alpha(i):
> for j from 2 to M do
> if (j+(i-2)=M) then g(i):=q(i); break; else
g(i):=q(i).alpha(j+(i-2)+1);q(i):=g(i): end if
> od:

> if (i-1=1) then G1(i-1):= g(i) else G1(i-1):= beta(i-
1).g(i) end if

> od:

> G1(M):=beta(M):
> G(1):=G1(1):
> for i from 1 to M-1 do
> G(i+1):= <G(i),G1(i+1)>:
> od:

> rtable_elems(G(M)):

```

```

> D1:= G(M).R.MatrixInverse(Qm).MatrixInverse(G(M)):
> X:= Vector(2^(M+1)+1):
> for i from 2 to 2^(M+1) do
> X(i):= (2*(i-1)-1)/(2^(M+2)-2):
> od:
> X(1):=0:X(2^(M+1)+1):=1:
> X;

> chi:=(i,j,x)-> piecewise((j/2^i)<=x and x<=((j+1)/2^i),
2^i*x-j, ((j+1)/2^i)<= x and x<= ((j+2)/2^i), 2-(2^i*x-
j), 0 ):
> for i from 2 to M do
> chi(i,-1,x) :=piecewise(0 <= x and x <= 1/2^i, 2-(2^i*x-
(-1)), 0 ):
> chi(i,(2^i)-1,x) := piecewise((((2^i)-1)/2^i)<=x and
x<=(((2^i)-1+1)/2^i), 2^i*x-((2^i)-1), 0)
> od:
> mu:= (i,j,x)-> 1/6* piecewise((j/2^i)<=x and x<=((j+
1/2)/2^i), 2^i*x-j, ((j+ 1/2)/2^i)<=x and x<= ((j+1)/2^i), 4-
7*(2^i*x-j), ((j+1)/2^i) <= x and x <= ((j+3/2)/2^i), -
19+16*(2^i*x-j), ((j+3/2)/2^i)<= x and x <= ((j+2)/2^i), 29-
16*(2^i*x-j), ((j+2)/2^i)<= x and x <= ((j+5/2)/2^i), -
17+7*(2^i*x-j), ((j+5/2)/2^i)<= x and x <= ((j+3)/2^i), 3-
(2^i*x-j), 0 ):
> chi(2,2,9/14);
> for i from 2 to M do
> mu(i,-1,x) :=piecewise(0 <= x and x <= (1/2)/2^i,-6
+23*(2^i*x), ((1/2)/2^i) <= x and x <= ((1)/2^i), 14 -
17*(2^i*x), ((1)/2^i) <= x and x <= ((3/2)/2^i), -10
+7*(2^i*x), ((3/2)/2^i) <= x and x <= ((2)/2^i), 2-(2^i*x), 0
):
> mu(i,((2^i)-2),x) := piecewise((((2^i)-2)/2^i)<=x and
x<=(((2^i)-2)+ 1/2)/2^i), 2-(((2^i)-2)+2-2^i*x), (((2^i)-
2)+ 1/2)/2^i)<=x and x<= (((2^i)-2)+1)/2^i), -
10+7*(((2^i)-2)+2-2^i*x), (((2^i)-2)+1)/2^i) <= x and x <=
(((2^i)-2)+3/2)/2^i), -14-17*(((2^i)-2)+2-2^i*x), (((2^i)-
2)+3/2)/2^i)<= x and x <= (((2^i)-2)+2)/2^i), -
6+23*(((2^i)-2)+2-2^i*x), 0)

```

```

> od:
> mu(2,-1,1/14);
#####33 Integration
> h:=1/4:
> for i from 0 to 4 do
> t(i):= i*h:
> od:
> t(4);
> mut:=Matrix(5,5):
> Qs:= Vector(2^(M+1)+1):
> for i from 1 to 5 do
> for j from 1 to 5 do
> mut(i,j):= (chi(2,i-2,t(j-1))):
> od:
> mut1:=Matrix(2^(M+1)+1-5,5):
> od:
> for k from 2 to M do
> #for i from 0 to (2^(M)-4) do
> for l from 1 to 2^k do
> for j from 1 to 5 do
> mut1((2^k-4)+l,j):= (mu(k,l-2,t(j-1))):
> od:
> od:
> od:
> #od:
> mutt:=<mut,mu1>:
> for i from 1 to 5 do

```

```

> KE(i-1):=3*x*t(i-1)*mutt(1..2^(M+1)+1,i):
> od:
> KE(1):
intg:=(2*h/45)*(7*KE(0)+32*KE(1)+12*KE(2)+32*KE(3)+7*KE(4))
:
> intgr:=Matrix(2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> for j from 1 to 2^(M+1)+1 do
> intgr(i,j):=subs(x=X(j),intg(i)):
> od:
> od:
> intgr:
> muX:=Matrix(5,2^(M+1)+1):
> for i from 1 to 5 do
> for j from 1 to 2^(M+1)+1 do
> muX(i,j):=(chi(2,i-2,X(j))):
> od:
> od:
> muX:
  > muX1:=Matrix(2^(M+1)+1-5,2^(M+1)+1):
> for k from 2 to M do
> for l from 1 to 2^k do
> for j from 1 to 2^(M+1)+1 do
> muX1((2^k-4)+1,j):=(mu(k,l-2,X(j))):
> od:
> od:
> od:
> muX1:

```

```

> muXX:=<muX,muX1>:
> #####3 The solution
> Z:=D1.muXX:
> ZZ:=Z-intgr:
> F:=Matrix(1,2^(M+1)+1): C:=Matrix(1,2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> F(1,i):= evalf(3*exp(3*X(i))-(1/3)*(2*exp(3)+1)*X(i));
> od:
> ZZ(1..2^(M+1)+1,1):=muXX(1..2^(M+1)+1,1):
> F(1,1):=1:
> C:=F.MatrixInverse(ZZ):
> U:=(C.muXX):
> u2:=x->evalf(exp(3*x)):
> u:=Transpose(U):
> u1:=convert(u,Vector):
> r:=plot( X,u1,color="Blue",style = point, symbol =
asterisk, symbolsize =15,legend = ["Numerical solu" ]):
r2:=plot(u2(x),x=0..1,color="Red",thickness=3,legend =
["Exact solu" ]):
> display(r,r2);
> ##### error
> Ue:=Vector(2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> Ue(i):=(u2(X(i)));
> od:
> Erroru:=Vector(2^(M+1)+1):
> Erroru:=abs(Ue-u1):

```

```

> A2:=plot( X, Erroru,x =
0..1,color="Blue",thickness=3,legend = ["E r r o r =| UE -
Un |" ]

> ):

> display(A2,titlefont = ["COURIER", 15],labelfont =
["COURIER", 10],axesfont = ["COURIER", 8]);

> X1:= Vector(11):

> for i from 1 to 11 do

> X1(i):= (i-1)*1/10:

> od:

> evalf(X1(11));

> muX2:=Matrix(5,11):

> for i from 1 to 5 do

> for j from 1 to 11 do

> muX2(i,j):= (chi(2,i-2,X1(j))):

> od:

> od:

> muX2:

> muX3:=Matrix(2^(M+1)+1-5,11):

> for k from 2 to M do

> for l from 1 to 2^k do

> for j from 1 to 11 do

> muX3((2^k-4)+l,j):= (mu(k,l-2,X1(j))):

> od:

> od:

> od:

> muX3:

> muXX2:=<muX2,muX3>:

> U2:=C.muXX2:

```

```

> rtable_elems(U2):
> NUM:= Matrix (10,1):Exact:= Matrix (10,1):Error:= Matrix
(10,1):
> for i from 1 to 10 do
> NUM(i,1):= U2(1,i+1):
> Exact(i,1):=u2(X1(i+1)):
> Error(i,1):=abs(Exact(i,1)-NUM(i,1))
> od:
evalf(X1(2..11));exact;Exact;Nume;(NUM);errorr;Error;

```

Maple Code for Homotopy Perturbation Method for Example 4.1

```

> restart;
> with(LinearAlgebra):with(plots):with(PDEtools):
> # This is a program to solve the integro-differential
equation on the form:
> Diff(v(x),x)=f(x)+ Int(G(x,y)*v(y),y=0..1);
> a:=1;
> f:= 3*exp(3*x)-(1/3)*(2*exp(3)+1)*x;
> G:= 3*x*y;
> n:=4;
> wm1[0]:= (int(f,x)):
> wm[0]:=y->subs(x=y,wm1[0]):
> c:=a-eval(wm[0](0)):
> ws[0]:= wm1[0]+c:
> w[0]:=subs(x=y,ws[0]):
> for i from 1 to n do
> wm[i]:= int(G*w[i-1],y=0..1);
> ws[i]:= (int(wm[i],x));
> w[i]:=subs(x=y,ws[i]);

```

```

> od:

> u:=combine(add(ws[k],k=0..n));

> um:= y-> subs(x=y,u):

> uexact:= x-> exp(3*x):

> r:=plot( um(x),x=0..1,color="Blue",style = point, symbol
= asterisk, symbolsize =15,legend = ["Numerical solu" ]):

r2:=plot(uexact(x),x=0..1,color="Red",thickness=3,legend =
["Exact solu" ]):

> display(r,r2);

> Erroru:=x->abs(uexact(x)-um(x)):

> Erroru(1);

> A2:=plot( Erroru(x),x =
0..1,color="Blue",thickness=3,legend = ["E r r o r =| UE -
Un |" ]):

> display(A2,titlefont = ["COURIER", 15],labelfont =
["COURIER", 10],axesfont = ["COURIER", 8]);

> X1:= Vector(10):NUM:= Matrix (10,1):Exact:= Matrix
(10,1):Error:= Matrix (10,1):

> for i from 1 to 10 do

> X1(i):= i*1/10:

> od:

  > for i from 1 to 10 do

> NUM(i,1):= evalf(um(X1(i))):

> Exact(i,1):=evalf(uexact(X1(i))):

> Error(i,1):=evalf(abs(Exact(i,1)-NUM(i,1)))

> od:

evalf(X1(2..11));exact;Exact;Nume;(NUM);errorr;Error;

```

Maple Code for B-spline scaling functions and wavelets method for example 4.2

```

> restart:

> with(plots):with(LinearAlgebra):

```

```

> M:=6:
> 2^(M+1)+1;
> #####3 Q & R
> Q:=Matrix(2^(M+1)+1):R:=Matrix(2^(M+1)+1):
> for i from 1 to (2^(M+1)+1) do
> for j from 1 to (2^(M+1)+1) do
> if (i=j+1) then Q(i,j):= 1/24; R(i,j):= 1/2 elif (j=i+1)then Q(i,j):= 1/24;R(i,j):= -1/2 elif
(i=j) then Q(i,j):= 1/6 end if
> od;
> od:
> Q(1,1):=1/12:Q(2^(M+1)+1,2^(M+1)+1):=1/12:R(1,1):=-
1/2:R(2^(M+1)+1,2^(M+1)+1):=1/2:
> Qm:= 1/2^(M-1)*Q:
> for j from 2 to M do
> alpha(j):= Matrix(2^j+1,2^(j+1)+1):
> beta(j):=Matrix(2^j,2^(j+1)+1):
> od:
> for k from 2 to M do
> for i from 1 to 2^k+1 do
> for j from 1 to 2^(k+1)+1 do
> if (j=2*i-1) then alpha(k)(i,j):= 1 elif (j=2*(i-1))then alpha(k)(i,j):= 1/2 elif (j=2*i)
then alpha(k)(i,j):= 1/2 end if
> od:
> od:
> od:
> for k from 2 to M do
> for i from 2 to 2^k-1 do
> for j from 1 to 2^(k+1)+1 do

```

```

> if (j=2*i-1) then beta(k)(i,j):= -1/2 elif (j=2*(i-1))then beta(k)(i,j):= 1/12 elif (j=2*i)
then beta(k)(i,j):= 5/6 elif (j=2*i+1) then beta(k)(i,j):= -1/2 elif (j=2*(i+1)) then
beta(k)(i,j):=1/12 end if

> od:

> od:

> beta(k)(1,1):= -1:beta(k)(1,2):= 11/12:beta(k)(1,3):= -1/2:beta(k)(1,4):=
1/12:beta(k)(2^k,2^(k+1)+1):= -1:beta(k)(2^k,2^(k+1)):= 11/12:beta(k)(2^k,2^(k+1)-
1):= -1/2:beta(k)(2^k,2^(k+1)-2):= 1/12:

> od:

> for i from 2 to M do

> q(i):=alpha(i):

> for j from 2 to M do

> if (j+(i-2)=M) then g(i):=q(i); break; else g(i):=q(i).alpha(j+(i-2)+1);q(i):=g(i): end if

> od:

> if (i-1=1) then G1(i-1):= g(i) else G1(i-1):= beta(i-1).g(i) end if

> od:

> G1(M):=beta(M):

> G(1):=G1(1):

> for i from 1 to M-1 do

> G(i+1):= <G(i),G1(i+1)>:

> od:

> rtable_elems(G(M)):

> #####3 D

> D1:= G(M).R.MatrixInverse(Qm).MatrixInverse(G(M)):

> X:= Vector(2^(M+1)+1):

> for i from 2 to 2^(M+1) do

> X(i):= (2*(i-1)-1)/(2^(M+2)-2):

> od:

> X(1):=0:X(2^(M+1)+1):=1:

```

```

> X;

> chi:=(i,j,x)-> piecewise ((j/2^i)<=x and x<=((j+1)/2^i), 2^i*x-j,((j+1)/2^i)<= x and
x<= ((j+2)/2^i),2 -(2^i*x-j),0 ):

> for i from 2 to M do

> chi(i,-1,x) :=piecewise(0 <= x and x <= 1/2^i,2 -(2^i*x-(-1)),0 ):

> chi(i,(2^i)-1,x) := piecewise((((2^i)-1)/2^i)<=x and x<=(((2^i)-1+1)/2^i),2^i*x-
((2^i)-1) ,0)

> od:

> mu:=(i,j,x)-> 1/6* piecewise((j/2^i)<=x and x<=((j+ 1/2)/2^i),2^i*x-j,((j+
1/2)/2^i)<=x and x<= ((j+1)/2^i), 4-7*(2^i*x-j), ((j+1)/2^i) <= x and x <= ((j+3/2)/2^i),
-19+16*(2^i*x-j),((j+3/2)/2^i)<= x and x <= ((j+2)/2^i), 29-16*(2^i*x-j),((j+2)/2^i)<=
x and x <= ((j+5/2)/2^i), -17+7*(2^i*x-j), ((j+5/2)/2^i)<= x and x <= ((j+3)/2^i), 3-
(2^i*x-j),0 ):

> chi(2,2,9/14);

> for i from 2 to M do

> mu(i,-1,x) :=piecewise(0 <= x and x <= (1/2)/2^i,-6 +23*(2^i*x),((1/2)/2^i) <= x and
x <= ((1)/2^i),14 -17*(2^i*x),((1)/2^i) <= x and x <= ((3/2)/2^i),-10
+7*(2^i*x),((3/2)/2^i) <= x and x <= ((2)/2^i),2 -(2^i*x),0 ):

> mu(i,((2^i)-2),x) := piecewise((((2^i)-2)/2^i)<=x and x<=(((2^i)-2)+ 1/2)/2^i),2-
(((2^i)-2)+2-2^i*x),((((2^i)-2)+ 1/2)/2^i)<=x and x<= (((2^i)-2)+1)/2^i), -10+7*(((2^i)-
2)+2-2^i*x), (((2^i)-2)+1)/2^i) <= x and x <= (((2^i)-2)+3/2)/2^i), -14-17*(((2^i)-
2)+2-2^i*x),((((2^i)-2)+3/2)/2^i)<= x and x <= (((2^i)-2)+2)/2^i), -6+23*(((2^i)-2)+2-
2^i*x),0)

> od:

> mu(2,-1,1/14);

> #####33 Integration

> h:=1/8:

> for i from 0 to 8 do

> t(i):= i*h:

> od:

> t(8);

> mut:=Matrix(5,9):

> Qs:= Vector(2^(M+1)+1):

```

```

> for i from 1 to 5 do
> for j from 1 to 9 do
> mut(i,j):= (chi(2,i-2,t(j-1))):
> od:
> mut1:=Matrix(2^(M+1)+1-5,9):
> od:
> for k from 2 to M do
> #for i from 0 to (2^(M)-4) do
> for l from 1 to 2^k do
> for j from 1 to 9 do
> mut1((2^k-4)+l,j):= (mu(k,l-2,t(j-1))):
> od:
> od:
> od:
> #od:
> mutt:=<mut,mu1>:
> for i from 1 to 9 do
> KE(i-1):=evalf(2*Pi*sin(4*Pi*x+2*Pi*t(i-1))*mutt(1..2^(M+1)+1,i)):
> od:
> KE(1):
> intg:=(4*h/14175)*(989*KE(0)+5888*KE(1)-928*KE(2)+10496*KE(3)-
4540*KE(4)+10469*KE(5)-928*KE(6)+5888*KE(7)+989*KE(8)):
> intgr:=Matrix(2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> for j from 1 to 2^(M+1)+1 do
> intgr(i,j):=subs(x=X(j),intg(i)):
> od:

```

```

> od:
> intgr:
> muX:=Matrix(5,2^(M+1)+1):
> for i from 1 to 5 do
> for j from 1 to 2^(M+1)+1 do
> muX(i,j):= (chi(2,i-2,X(j))):
> od:
> od:
> muX:
> muX1:=Matrix(2^(M+1)+1-5,2^(M+1)+1):
> for k from 2 to M do
> for l from 1 to 2^k do
> for j from 1 to 2^(M+1)+1 do
> muX1((2^k-4)+l,j):= (mu(k,l-2,X(j))):
> od:
> od:
> od:
> muX1:
> muXX:=<muX,muX1>:
> #####3 The solution
> Z:=D1.muXX:
> ZZ:=Z-intgr:
> F:=Matrix(1,2^(M+1)+1): C:=Matrix(1,2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> F(1,i):= evalf(-2*Pi*sin(2*Pi*X(i))-Pi*sin(4*Pi*X(i)));
> od:
> ZZ(1..2^(M+1)+1,1):=muXX(1..2^(M+1)+1,1):

```

```

> F(1,1):=1:
> C:=F.MatrixInverse(ZZ):
> U:=(C.muXX):
> u2:=x->evalf(cos(2*Pi*x)):
> u:=Transpose(U):
> u1:=convert(u,Vector):
> r:=plot( X,u1,color="Blue",style = point, symbol = asterisk, symbolsize =15,legend =
["Numerical solu" ]):
> r2:=plot(u2(x),x=0..1,color="Red",thickness=3,legend = ["Exact solu" ]):
> display(r,r2);
> ##### error
> Ue:=Vector(2^(M+1)+1):
> for i from 1 to 2^(M+1)+1 do
> Ue(i):=(u2(X(i)));
> od:
> Erroru:=Vector(2^(M+1)+1):
> Erroru:=abs(Ue-u1):
> A2:=plot( X, Erroru,x = 0..1,color="Blue",thickness=3,legend = ["E r r o r =| UE - Un
|" ]
> ):
> display(A2,titlefont = ["COURIER", 15],labelfont = ["COURIER", 10],axesfont =
["COURIER", 8]);
> X1:= Vector(11):
> for i from 1 to 11 do
> X1(i):= (i-1)*1/10:
> od:
> evalf(X1(11));
> muX2:=Matrix(5,11):

```

```

> for i from 1 to 5 do
> for j from 1 to 11 do
> muX2(i,j):= (chi(2,i-2,X1(j))):
> od:
> od:
> muX2:
> muX3:=Matrix(2^(M+1)+1-5,11):
> for k from 2 to M do
> for l from 1 to 2^k do
> for j from 1 to 11 do
> muX3((2^k-4)+1,j):= (mu(k,l-2,X1(j))):
> od:
> od:
> od:
> muX3:
> muXX2:=<muX2,muX3>:
> U2:=C.muXX2:
> rtable_elems(U2):
> NUM:= Matrix (10,1):Exact:= Matrix (10,1):Error:= Matrix (10,1):
> for i from 1 to 10 do
> NUM(i,1):= U2(1,i+1):
> Exact(i,1):=u2(X1(i+1)):
> Error(i,1):=abs(Exact(i,1)-NUM(i,1))>
> od:
> evalf(X1(2..11));exact;Exact;Nume;(NUM);errorr;Error;

```

Maple Code for Homotopy Perturbation Method for Example 4.2

```

> restart;

```

```

with(LinearAlgebra):with(plots):with(PDEtools):

> # This is a program to solve the integro-differential
equation on the form:

> Diff(v(x),x)=f(x)+ Int(G(x,y)*v(y),y=0..1);

> a:=1;

> f:= -2*Pi*sin(2*Pi*x)-Pi*sin(4*Pi*x);

> G:= 2*Pi*sin(4*Pi*x+2*Pi*y);

> n:=8;

> wm1[0]:= ( int(f,x)):

> wm[0]:=y->subs(x=y,wm1[0]):

> c:=a-eval(wm[0](0)):

> ws[0]:= wm1[0]+c:

> w[0]:=subs(x=y,ws[0]):

> for i from 1 to n do

> wm[i]:= int(G*w[i-1],y=0..1);

> ws[i]:= (int(wm[i],x));

> w[i]:=subs(x=y,ws[i]);

> od:

> u:=combine(add(ws[k],k=0..n));

> um:= y-> subs(x=y,u):

> uexact:= x-> cos(2*Pi*x):

> r:=plot( um(x),x=0..1,color="Blue",style = point, symbol
= asterisk, symbolsize =15,legend = ["Numerical solu" ]):

r2:=plot(uexact(x),x=0..1,color="Red",thickness=3,legend =
["Exact solu" ]):

> display(r,r2);

> Erroru:=x->abs(uexact(x)-um(x)):

> A2:=plot( Erroru(x),x =
0..1,color="Blue",thickness=5,legend = ["E r r o r =| u-uM
|" ]):

```

```

> display(A2,titlefont = ["COURIER", 15],labelfont =
["COURIER", 10],axesfont = ["COURIER", 8]);

> X1:= Vector(11):NUM:= Matrix (10,1):Exact:= Matrix
(10,1):Error:= Matrix (10,1):

> for i from 1 to 11 do

> X1(i):= (i-1)*1/1:

> od:

  > for i from 1 to 10 do

> NUM(i,1):= evalf(um(X1(i))):

> Exact(i,1):=evalf(uexact(X1(i))):

> Error(i,1):=evalf(abs(Exact(i,1)-NUM(i,1)))

> od:

evalf(X1(2..11));exact;Exact;Nume;(NUM);errorr;Error;

```

Matlab Code for Legendre Polynomial Method

```

%%

clc

clear

close

%%

syms x t;

a = -1;

b = 1;

N = 8;

Lambda = 1;

f=[-x x 1];% F0 F1 F2

gx= exp(x)- 2*sin(x);

kxt= sin(x)*exp(-t);

%% xi

```

```

    xi=zeros(1,N+1);

for i=0:N
    xi(1,i+1) = a+i*((b-a)/N);
end

%% F
F=zeros(N+1,N+1,N);
for k=1:length(f)
    for i=0:N
        F(i+1,i+1,k) = subs(f(k),{x},{xi(i+1)});
    end
end

%% P
Pk=zeros(N);
for j = 0:N
    for i = 0:N
        P=legendrep(i, 'X');
        X=xi(j+1);
        Pk(j+1,i+1)=eval(P);
    end
end

%% PI
PI = legendrePi(N+1);

%% KXT
Kl = legendreKT(N,kxt);

%% Q
n=0:N;
Q = diag(2./(2*n+1));

```

```

%% G
G = subs(gx,x,xi);
G=double(G);
%% W
WD=zeros(N+1,N+1,N-1);
for k=0:N-1
    WD(:, :, k+1)=F(:, :, k+1)*Pk*((PI')^k);
end
%% W
WX=zeros(N+1);
for k=0:N-1
    WX(:, :)=WX(:, :)+WD(:, :, k+1);
end
W=WX-Lambda*Pk*Kl*Q;
%% U
U0=zeros(1,N);
for i = 0:N
    P=legendrep(i, 'x');
    X=0;
    U0(1,i+1)=eval(P);
end
%%
U1=zeros(1,N);
U1(1,1)=0;
for i = 1:N
    P=legendrep(i, 'x');
    P=eval(P);

```

```

P=diff(P);
U1(1,i+1)=subs(P,x,0);
end
%%
WU=[W(1:N-1,:);U0;U1];
GU=[G(1:N-1),1,1];
%%
A=WU\GU';
%%
Y='0';
for i = 0:N
    aux= ['(' num2str(A(i+1)) '*' legendrep(i,'x') ')'];
    Y=[Y '+' aux];
end
Y=eval(Y);
%% results
xu=-1:0.2:1;
Aprx_sol_legndr=double(subs(Y,{x},xu));
exct_sol=exp(xu);
legndr_err=abs(exct_sol-Aprx_sol_legndr);
%%
load pd.mat
%% plot
figure(1)
% Create multiple lines using matrix input to plot
plot1 = plot(xu,exct_sol,xu,Aprx_sol_legndr);
set(plot1(1),'Marker','*','LineWidth',2);

```

```
set(plot1(2), 'Marker', 'square', 'LineWidth', 1, 'Color', [1 0
0]);
```

```
grid
```

```
legend('exct', 'Aprx legndr')
```

```
figure(2)
```

```
plot(xu, legndr_err)
```

```
grid
```

```
legend('legndr err')
```

Note this part explains how to calculate Legendre polynomials

```
function p=legendrep(m,x)
```

```
% function which construct Legendre polynomial Pm(x)
```

```
% where M is the degree of polynomial and X is the
variable.
```

```
% Result - P is the char string that should be evaluated
EVAL(P)
```

```
% Example:
```

```
% P=legendrep(2,'cos(theta)') will produce
```

```
% P='(3*cos(theta).^2 -1)/2' which then can be evaluated
as
```

```
% theta=0.3; P=legendrep(2,'cos(theta)'); Lp=eval(P);
produce
```

```
% Lp = 0.8690
```

```
% For Matlab R14 the following example can be used:
```

```
% x=-5:.1:5; p=legendrep(5,'x .*cos(x)'); xp = eval(p);
```

```
% figure; plot(x, xp, 'r.-'); grid
```

```
%% References:
```

```
% Gradshteyn, Ryzhik "Table of Integrals Series and
Products", 6th ed., p.973
```

```
%
```

```
%
```

```

% Sergei Koptenko, Resonant Medical Inc., Toronto
% sergei.koptenko@resonantmedical.com
% _____March/30/2004_____
%%

switch m
case 0
    p='1';    return

case 1
    p=x;     return

case 2
    p=['(3*' x '^2 -1)/2'];    return

case 3
    p=['(5*' x '^3 - 3 *' x ') /2'];    return

case 4
    p=['(35 *' x '^4 - 30 *' x '^2 + 3)/8'];    return

case 5
    p=['( 63 *' x '^5 - 70 *' x '^3 + 15 *' x ') /8'];
return

case 6
    p=['(231 *' x '^6 - 315 *' x '^4 + 105 *' x '^2 -
5)/16'];    return

    case 7
        p=['(429 *' x '^7 - 693 *' x '^5 +315 *' x '^3 -35
*' x ') /16'];    return

case 8
    p=['(6435 *' x '^8 -12012 *' x '^6 + 6930 *' x '^4
-1260 *' x '^2 +35)/128'];    return

case 9
    p=['(12155 *' x '^9 -25740 *' x '^7 +18018 *' x
'^5 -4620 *' x '^3 +315 *' x ') /128'];

```

```

    return

case 10
    p=['(46189 * ' x '^10 -109395 * ' x '^8 +90090 * ' x
'^6 -30030 * ' x '^4 +3465 * ' x '^2 -63)/256'];

    return

case 11
    p=['(88179 * ' x '^11 -230945 * ' x '^9 +218790 * ' x
'^7 -90090 * ' x '^5 +15015 * ' x '^3 -693 * ' x '
)/256'];

    return

case 12
    p=['(676039 * ' x '^12 -1939938 * ' x '^10 +2078505 * '
x '^8 -1021020 * ' x '^6 +225225 * ' x '^4 -18018 * ' x
'^2 +231)/1024'];

    return

otherwise
    iii=m-10;    %shift counter

    pp=strvcat(legendrep(11,x),legendrep(12,x)); % get last
two members

    for ii=3:1:iii,    % Begin construct from 13th member
        p_ii=[num2str(1/(ii)) ' * (' num2str(2*ii-1) ' * '
x '.*(' ...

        deblank(pp(ii-1,:)) ')-' num2str(ii-1) '.*('
deblank(pp(ii-2,:)) '))'];

        pp=strvcat(pp, p_ii);

    end

p=deblank(pp(iii,:)); % remove trailing blanks

    return

end

return

```

Matlab Code for Taylor collocation method

```

clear all

clc

close all

format long

%%

%*****

%*** INPUT DATA

%*****

syms x t

%%

a=-1;

b=1;

c=0;

N=5;

K=2;

Lambda=1;

FX=exp(x)- 2*sin(x);

KXT=sin(x)*exp(-t);

PS=[1,-x, x];

%% xi

xi=zeros(1,N+1);

for i=0:N

    xi(1,i+1)=a+i*((b-a)/N);

end

%% H

H=zeros(N+1);

for n=0:N

```

```

    for m=0:N
        expnt=n+m+1;
        H(n+1,m+1)=(b-c)^expnt-(a-c)^expnt/expnt;
    end
end

%% C
C=zeros(N+1);
for n=0:N
    for m=0:N
        C(n+1,m+1)=(xi(1,n+1))^m;
    end
end

%% F
F=zeros(N+1,1);
for i=0:N
    F(i+1,1)=subs(FX,{x},{xi(i+1)});
end

%% K
KL=zeros(N+1);
for n=0:N
    for m=0:N
        for td=0:m
            if td > 0
                PK=diff(KXT,t);
                KXT=PK;
            end
        end
    end
end

```

```

end
for xd=0:n
    if xd > 0
        PK=diff(KXT,x);
        KXT=PK;
    end
end

KL(n+1,m+1)=(1/(factorial(n)*factorial(m)))*subs(KXT,{x,t},
{0,0});

KXT=sin(x)*exp(-t);

end
end

%% M's

M=zeros(N+1,N+1,K);

fctrl=0;
for k=0:K
    fctrl=0;
    for n=0:N
        for m=(k):N
            if n==(m-k)
                M(n+1,m+1,(k+1))=1/factorial(fctrl);
                fctrl=fctrl+1;
            end
        end
    end
end

end

%% P's

```

```

% P0=-x
% P1=x
% P2=1
% PS=[1,x,-x];
P=zeros(N+1,N+1,K);
for k=0:K
    for n=0:N
        for m=0:N
            if n==m
                P(n+1,m+1,(k+1))=subs(PS(k+1),{x},{xi(m+1)});
            end
        end
    end
end

%% YS
YS=zeros(N+1,1);
YS(1)=1;
YS(2)=1;

%% FU
FU=[F ; YS];

%% U
U=zeros(N+1);
for n=0:N
    for m=0:N
        if ((n==m) && (YS(m+1)==1))
            U(n+1,m+1)=1;
        end
    end
end

```

```

        end
    end
end

%% W
WD=zeros(N+1,N+1,K);
for k=0:K
    WD(:,:,k+1)=P(:,:,k+1)*C*M(:,:,k+1);
end

WX=zeros(N+1);
for k=0:K
    WX(:,:,k)=WX(:,:,k)+WD(:,:,k+1);
end

W=WX-Lambda*C*KL*H*M(:,:,1);

%% WU
WU=[W ; U];

%% A
A=WU\FU;

%% paper
Co=zeros(1,N+1);
for n =0:N
    Co(1,n+1) = (A(n+1)/factorial(n));
end

%% Y
YT=0;
Co=[1 1 0.500343 0.166886 0.0403378 0.00577493];
for n =0:N
    YT = YT + Co(n+1)*(x - c)^n;

```

```
end

%% Plot
xu=-1:0.2:1;
Aprx_sol_tyl=double(subs(YT,{x},xu));
exct_sol=exp(xu);
figure(2)
plot(xu,exct_sol,':',xu,Aprx_sol_tyl,'g')
grid
legend('exct','Aprx tyl')

%% Table
% T =
table(xu',exct_sol',Aprx_sol_tyl','VariableNames',{'X'
'exct_sol' 'Aprx_sol'});

% filename = 'DataTable.xlsx';

% writetable(T,filename,'Sheet',1)

%% err
figure(4)
tyl_err=abs(exct_sol-Aprx_sol_tyl);
plot(xu,tyl_err)
grid
legend('tylr err')

%%tbl
tbl_tyl=[xu ; exct_sol ; Aprx_sol_tyl ; tyl_err]';
```

جامعة النجاح الوطنية

كلية الدراسات العليا

معادلة فريدهولم الخطية التكاملية التفاضلية من النوع الثاني

إعداد

خلود نضال اسعيد ظاهر

إشراف

أ.د. ناجي قطناني

قدمت هذه الأطروحة استكمالاً لمتطلبات الحصول على درجة الماجستير في الرياضيات بكلية الدراسات العليا، في جامعة النجاح الوطنية ، نابلس – فلسطين.

2016

ب

معادلة فريدهولم الخطية التكاملية التفاضلية من النوع الثاني

إعداد

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الملخص

في هذه الاطروحة ركزنا على حل معادلة فريدهولم التكاملية التفاضلية من النوع الثاني لان لها مجال واسع في التطبيقات الفيزيائية. وقمنا باستقصاء بعض الطرق التحليلية والعددية لحل هذه المعادلة. الطرق التحليلية شملت: طريقة الحساب المباشر، طريقة التكرار التغييري، طريقة أدومين التحليلية، طريقة أدومين التحليلية المعدلة، ظاهرة الضوضاء وطريقة حل السلسلة.

الطرق العددية التي تناولناها هي: طريقة بي سبلاين و وظائف قياس المويجات، طريقة هوموتوبي الاضطرابية، طريقة كثيرات الحدود و دوال لجندر و طريقة تايلور التجميعية. وبشكل خاص الأمثلة العددية التي تناولناها نفذت باستخدام هذه الطرق العددية لحل معادلة فريدهولم التكاملية التفاضلية الخطية من النوع الثاني. تم وضع مقارنة بين هذه الطرق العددية حيث أظهرت لنا النتائج العددية أن طريقة هوموتوبي الاضطرابية وطريقة كثيرات الحدود ودوال لجندر أنها الأكثر كفاءة بالمقارنة مع الطرق العددية الأخرى وذلك بناء على الأمثلة التي استخدمناها.