# - <br> <br> University <br> <br> University of Cyprus 

 of Cyprus}

# DEPARTMENT OF MATHEMATICS AND STATISTICS 

Numerical Solution of Fredholm Integral and Integro-Differential Equations of the second type

Maria Charalambous

# 4 <br> <br> University <br> <br> University of Cyprus 

 of Cyprus}

## DEPARTMENT OF MATHEMATICS AND STATISTICS

# Numerical Solution of Fredholm Integral and Integro-Differential Equations of the second type 

Maria Charalambous<br>Advisor: Christos Xenophontos

A Dissertation Submitted to the University of Cyprus in Partial Fulfillment of the Requirements for the Degree of Master of Philosophy
© Copyright by Maria Charalambous, 2023

## Contents

1 Introduction ..... 1
2 Integral Equations ..... 2
3 Numerical Solution of Fredholm Integral Equations of the $2^{\text {nd }}$ type ..... 5
4 Numerical Solution of Fredholm Integro-Differential Equations of the $2^{\text {nd }}$ type ..... 10
4.1 Singularly perturbed Fredholm Integro-Differential Equations of the $2^{\text {nd }}$ type ..... 12
5 Closing Remarks ..... 21

## 1 Introduction

In this thesis we study the numerical approximation of the solution to Fredholm integral and integro-differential equations of the second type, as these will be defined in the next chapter. Integral equations can be used to model population growth [5], biological systems [9], elastohydrodynamic lubrication [4], quantum scattering [3], and heat transfer [6]. Furthermore, ordinary and partial differential equations (ODEs and PDEs), which also arise in a wide range of physical problems, can be reformulated as integral equations. The advantage of the integral equation reformulation is that associated boundary and initial conditions (BCs and ICs) are incorporated within the integral equation, in contrast to ODEs and PDEs on which BCs and ICs are imposed.

It all began in 1823 when Abel proposed a generalization of the tautochrone problem whose solution involved the solution of an integral equation which was later designated as an integral equation of the first type, and in 1837 Liouville showed that the determination of a particular solution of a second order linear differential equation depends on solving an integral equation of a different type, called the integral equation of the second kind. After that, the work of Volterra, Fredholm, and Hilbert, starting in 1896, has made integral equations a discipline of its own, so to speak. In this thesis, we focus on Fredholm integral equations of the second type, and the approximation of their solution.

In Chapter 2, we give an overview of integral equations, including integrodifferential equations; we present their definition, the different categories they are divided in, existence and uniqueness of a solution, as well as some other general results.

In Chapter 3, we consider numerical approximations by means of classical composite quadrature rules and collocation, an approach called Nyström's method in the literature [2]. In particular, we utilize the composite Trapezoidal and Simpson's Rules for the approximation of the integral in the equation, in conjunction with collocation at the same nodal points.

In Chapter 4, we consider Fredholm integro-differential equations of the second type, and present the numerical approximation for their solution by the composite Trapezoidal rule, along with centered finite differences for the approximation of the derivative. In Section 4.1 we consider singularly perturbed Fredholm integro-differential equations of the second type, and present the numerical approximation for their solution, using the popular Shishkin mesh [8].

We close with some concluding remarks in Chapter 5.

## 2 Integral Equations

An equation that contains the unknown function $u(x)$ within an integral is called an integral equation. For example

$$
\begin{equation*}
f(x)=\int_{a}^{b} K(x, t) u(t) d t, x \in[a, b], \tag{1}
\end{equation*}
$$

where we assume that the so-called kernel $K(x, t)$ is a given function on the interval $[a, b] \times[a, b]$, satisfying

$$
\begin{equation*}
\int_{a}^{b} \int_{a}^{b}|K(x, t)|^{2} d x d t<\infty, \tag{2}
\end{equation*}
$$

and that $f(x)$ is a given, continuous function on $[a, b]$. Since the limits of integration are constant, (1) is called an integral equation of Fredholm type. The integral equation

$$
f(x)=\int_{a}^{x} K(x, t) u(t) d t, x \in[a, b],
$$

is of Voltera type, since it has a variable limit of integration. We will only consider Fredholm integral equations in this thesis.

If $u(x)$ appears inside and outside the integral, the Fredholm integral equation is called of the second type:

$$
\begin{equation*}
u(x)=f(x)+\int_{a}^{b} K(x, t) u(t) d t, x \in[a, b] . \tag{3}
\end{equation*}
$$

The celebrated Fredholm Alternative [2] tells us when an integral equation has a unique solution: If the homogeneous integral equation

$$
y(x)=\int_{a}^{b} K(x, t) y(t) d t, x \in[a, b],
$$

has only the zero solution $y(x)=0$, then the corresponding non-homogeneous integral equation

$$
y(x)=f(x)+\int_{a}^{b} K(x, t) y(t) d t, x \in[a, b],
$$

always has a unique solution. Conversely, if the homogeneous equation has nonzero solutions, then the non-homogeneous integral equation either has no solutions or has infinite solutions, and this depends on the given function $f(x)$ (see [2] for more details).

We mention that in the so-called case of a separable kernel, i.e. $K(x, t)=$ $k_{1}(x) k_{2}(t)$, for some smooth functions $k_{1}, k_{2}$, integral equations may be solved analytically, in most cases (see Example 2.1 below). However, in the more general case of a smooth kernel considered in this thesis, a numerical method is necessary. In Chapter 3 we will consider the numerical approximation of the solution to (3).

Example 2.1: Separable Kernel
We consider the problem

$$
y(x)=f(x)+\int_{a}^{b} k_{1}(x) k_{2}(t) y(t) d t, x \in[a, b],
$$

which may be written as

$$
y(x)=f(x)+k_{1}(x) \int_{a}^{b} k_{2}(t) y(t) d t .
$$

Multiplying by $k_{2}(x)$ and integrating from $a$ to $b$, gives

$$
\begin{array}{r}
\int_{a}^{b} k_{2}(x) y(x) d x=\int_{a}^{b} k_{2}(x) f(x) d x+\int_{a}^{b} k_{2}(x) k_{1}(x) d x \int_{a}^{b} k_{2}(t) y(t) d t \\
\left(1-\int_{a}^{b} k_{2}(x) k_{1}(x) d x\right) \int_{a}^{b} k_{2}(t) y(t) d t=\int_{a}^{b} k_{2}(x) f(x) d x
\end{array}
$$

from which we may calculate

$$
C^{*}:=\int_{a}^{b} k_{2}(t) y(t) d t=\frac{\int_{a}^{b} k_{2}(x) f(x) d x}{1-\int_{a}^{b} k_{2}(x) k_{1}(x) d x} \in \mathbb{R},
$$

provided

$$
1-\int_{a}^{b} k_{2}(x) k_{1}(x) d x \neq 0
$$

Hence, the solution is given by

$$
y(x)=f(x)+C^{*} k_{1}(x) .
$$

One other type of equation we will consider involves the derivative of the unknown function as well, e.g.,

$$
\begin{align*}
u^{\prime}(x) & =f(x)+\int_{a}^{b} K(x, t) u(t) d t, x \in(a, b],  \tag{4}\\
u(a) & =u_{a} \in \mathbb{R} . \tag{5}
\end{align*}
$$

This is an integro-differential equation (of first order in this case, since only first derivatives appear) of Fredholm type, and it includes the initial condition (5) in order to have a unique solution. We could have higher order integro-differential equations, but we will focus only on the type given by (4). In Section 4.1 we will consider the numerical approximation of the solution to (4)-(5).

A more general version of (4)-(5), not considered in this thesis, would be

$$
\begin{aligned}
u^{\prime}(x)+c(x) u(x) & =f(x)+\int_{a}^{b} K(x, t) u(t) d t, x \in(a, b], \\
u(a) & =u_{a} \in \mathbb{R}
\end{aligned}
$$

with, additionally, $c>0$ a given sufficiently smooth function. The methods presented in this thesis could easily be extended to this case with minor modifications.

## 3 Numerical Solution of Fredholm Integral Equations of the $2^{\text {nd }}$ type

Our goal in this chapter is to approximate the solution $u$ of (3). We begin by constructing a grid $a=x_{1}<\ldots<x_{m-1}<x_{m}=b$ and substitute each $x_{i}$ in equation (3), i.e. we use collocation:

$$
\begin{equation*}
u\left(x_{i}\right)=f\left(x_{i}\right)+\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t, i=1, \ldots, m \tag{6}
\end{equation*}
$$

The integral in (6) is approximated by numerical quadrature, based on the nodal points $x_{1}, x_{2}, \ldots, x_{m}$. Recall that in numerical integration, we use

$$
\int_{a}^{b} g(t) d t \approx \sum_{j=1}^{N} w_{j} g\left(t_{j}\right)
$$

where the $w_{j}$ are called weights and the $t_{j}$ nodes, ( $N$ is the number of terms). The different integration rules result from different choices for $w_{j}$ and $t_{j}$. We take $t_{j}=x_{i}, N=m$, when approximating the integral in (6), in order to have only the unknown values $u\left(x_{i}\right)$ without introducing new ones. This corresponds to the linear system

$$
\begin{equation*}
U_{i}=f\left(x_{i}\right)+\sum_{j=1}^{m} w_{j} K\left(x_{i}, x_{j}\right) U_{j}, i=1, \ldots m \tag{7}
\end{equation*}
$$

where $U_{i} \approx u\left(x_{i}\right)$. Solving the system results in the approximate values $u\left(x_{1}\right), \ldots, u\left(x_{m}\right)$. In matrix form, the linear system (7) is written as

$$
A \vec{U}=\vec{b},
$$

where

$$
[A]_{i, j}=\left[\begin{array}{cccc}
1-w_{1} K\left(x_{1}, x_{1}\right) & -w_{2} K\left(x_{1}, x_{2}\right) & \ldots & -w_{N} K\left(x_{1}, x_{m}\right)  \tag{8}\\
-w_{1} K\left(x_{2}, x_{1}\right) & 1-w_{2} K\left(x_{2}, x_{2}\right) & \ldots & -w_{N} K\left(x_{2}, x_{m}\right) \\
\vdots & \vdots & \ldots & \vdots \\
-w_{N} K\left(x_{m}, x_{1}\right) & -w_{2} K\left(x_{m}, x_{2}\right) & \ldots & 1-w_{N} K\left(x_{m}, x_{m}\right)
\end{array}\right]
$$

or $A=(I-M)$ with $I$ the identity matrix, $M_{i j}=w_{j} K\left(x_{i}, x_{j}\right)$, and

$$
\vec{b}=\left[\begin{array}{c}
f\left(x_{1}\right) \\
f\left(x_{2}\right) \\
f\left(x_{3}\right) \\
\vdots \\
f\left(x_{m}\right)
\end{array}\right] .
$$

The above method is known as Nyström's method in the literature [2].
For example, if we were to use the Composite Trapezoidal Rule as our numerical integration scheme, the weights $w_{j}$ and nodes $x_{j}$ in (7) are

$$
\begin{gathered}
\vec{w}=\left[w_{1}, \ldots, w_{m}\right]^{T}=\frac{1}{2}\left(\frac{b-a}{m}\right)[1,2 \ldots, 2,1], \\
x_{j}=\frac{b-a}{m} j, j=1, \ldots, m .
\end{gathered}
$$

The integration scheme is

$$
\int_{a}^{b} g(x) d x \approx \frac{1}{2}\left(\frac{b-a}{m}\right)\left[g\left(x_{1}\right)+2\left(g\left(x_{2}\right)+\ldots+g\left(x_{m-1}\right)\right)+g\left(x_{m}\right)\right]
$$

and the system matrix in (8) is

$$
[A]_{i, j}=\left[\begin{array}{ccccc}
1-K_{1,1} & -2 K_{1,2} & \ldots & -2 K_{1, m-1} & -K_{1, m}  \tag{9}\\
-K_{2,1} & 1-2 K_{2,2} & \ldots & -2 K_{2, m-1} & -K_{2, m} \\
\vdots & \vdots & \ldots & \vdots & \vdots \\
-K_{m, 1} & -2 K_{m, 2} & \ldots & -2 K_{m, m-1} & 1-K_{m, m}
\end{array}\right]
$$

where we used the notation $K_{i, j}=K\left(x_{i}, x_{j}\right)$.
Similarly, for the Composite Simpson's Rule, where the weights $w_{j}$ and nodes $x_{j}$ in (7) are

$$
\begin{gathered}
\vec{w}=\left[w_{1}, \ldots w_{m}\right]^{T}=\frac{1}{3}\left(\frac{b-a}{m}\right)[1,4,2,4, \ldots, 2,4,1], \\
x_{j}=\frac{b-a}{m} j, j=1, \ldots, m,
\end{gathered}
$$

the numerical integration scheme is

$$
\int_{a}^{b} g(x) d x \approx \frac{1}{3}\left(\frac{b-a}{m}\right)\left[g\left(x_{0}\right)+4 g\left(x_{1}\right)+2 g\left(x_{2}\right)+4 g\left(x_{3}\right)+\ldots+2 g\left(x_{m-2}\right)+4 g\left(x_{m-1}\right)+g\left(x_{m}\right)\right]
$$

and the system matrix in (8) is

$$
[A]_{i, j}=\left[\begin{array}{ccccccc}
1-K_{1,1} & -4 K_{1,2} & -2 K_{1,3} & \ldots & -2 K_{1, m-2} & -4 K_{1, m-1} & -K_{1, m}  \tag{10}\\
-K_{2,1} & 1-4 K_{2,2} & -2 K_{2,3} & \ldots & -2 K_{2, m-2} & -4 K_{2, m-1} & -K_{2, m} \\
\vdots & \vdots & \vdots & \ldots & \vdots & \vdots & \vdots \\
-K_{m, 1} & -4 K_{m, 2} & -2 K_{m, 3} & \ldots & -2 K_{m, m-2} & -4 K_{m, m-1} & 1-K_{m, m}
\end{array}\right]
$$

We illustrate the above in the following elementary example, with most calculations being displayed in detail.

## Example 3.1:

Let $a=0, b=1, f(x)=x^{2}, K(x, t)=e^{|x-t|}$, and divide [0, 1] into 4 subintervals, using the nodal points $x_{0}=0, x_{1}=0.25, x_{2}=0.5, x_{3}=0.75, x_{4}=1$. Using the Composite Trapezoidal Rule, we write the equations satisfied by $u\left(x_{i}\right)$, as
$u\left(x_{i}\right)=x_{i}^{2}+\frac{1}{2}\left(\frac{1-0}{4}\right)\left[e^{\left|x_{i}-x_{0}\right|} u\left(x_{0}\right)+2 e^{\left|x_{i}-x_{1}\right|} u\left(x_{1}\right)+2 e^{\left|x_{i}-x_{2}\right|} u\left(x_{2}\right)+2 e^{\left|x_{i}-x_{3}\right|} u\left(x_{3}\right)+e^{\left|x_{i}-x_{4}\right|} u\left(x_{4}\right)\right]$,
where $i=0,1,2,3,4$, or equivalently as

$$
\left[\begin{array}{ccccc}
\frac{7}{8} & -\frac{1}{4} e^{\frac{1}{4}} & -\frac{1}{4} e^{\frac{1}{2}} & -\frac{1}{4} e^{\frac{3}{4}} & -\frac{1}{8} e^{1} \\
-\frac{1}{8} e^{\frac{1}{4}} & \frac{3}{4} & -\frac{1}{4} e^{\frac{1}{4}} & -\frac{1}{4} e^{\frac{1}{2}} & -\frac{1}{8} e^{\frac{3}{4}} \\
-\frac{1}{8} e^{\frac{1}{2}} & -\frac{1}{4} e^{\frac{1}{4}} & \frac{3}{4} & -\frac{1}{8} e^{\frac{1}{4}} & -\frac{1}{8} e^{\frac{1}{2}} \\
-\frac{1}{8} e^{\frac{3}{4}} & -\frac{1}{4} e^{\frac{1}{2}} & -\frac{1}{4} e^{\frac{1}{4}} & \frac{3}{4} & -\frac{1}{8} e^{\frac{1}{4}} \\
-\frac{1}{8} e^{1} & -\frac{1}{4} e^{\frac{3}{4}} & -\frac{1}{4} e^{\frac{1}{2}} & -\frac{1}{4} e^{\frac{1}{4}} & \frac{7}{8}
\end{array}\right]\left[\begin{array}{c}
u\left(x_{0}\right) \\
u\left(x_{1}\right) \\
u\left(x_{2}\right) \\
u\left(x_{3}\right) \\
u\left(x_{4}\right)
\end{array}\right]=\left[\begin{array}{c}
0 \\
\frac{1}{16} \\
\frac{1}{4} \\
\frac{9}{16} \\
1
\end{array}\right] .
$$

Solving the system we find the approximate values of $u\left(x_{0}\right), u\left(x_{1}\right), u\left(x_{2}\right), u\left(x_{3}\right), u\left(x_{4}\right)$ as $u(0)=-1.2309, u(0.25)=-0.9738, u(0.5)=-0.6817, u(0.75)=-0.6118, u(1)=$ -0.4698 .

For the Composite Simpson's Rule, we have
$u\left(x_{i}\right)=x_{i}^{2}+\frac{1}{3}\left(\frac{1-0}{4}\right)\left[e^{\left|x_{i}-x_{0}\right|} u\left(x_{0}\right)+4 e^{\left|x_{i}-x_{1}\right|} u\left(x_{1}\right)+2 e^{\left|x_{i}-x_{2}\right|} u\left(x_{2}\right)+4 e^{\left|x_{i}-x_{3}\right|} u\left(x_{3}\right)+e^{\left|x_{i}-x_{4}\right|} u\left(x_{4}\right)\right]$,
where $i=0,1,2,3,4$, or equivalently as

$$
\left[\begin{array}{ccccc}
\frac{11}{12} & -\frac{1}{3} e^{\frac{1}{4}} & -\frac{1}{6} e^{\frac{1}{2}} & -\frac{1}{3} e^{\frac{3}{4}} & -\frac{1}{12} e^{1} \\
-\frac{1}{12} e^{\frac{1}{4}} & \frac{2}{3} & -\frac{1}{6} e^{\frac{1}{4}} & -\frac{1}{3} e^{\frac{1}{2}} & -\frac{1}{12} e^{\frac{3}{4}} \\
-\frac{1}{11} e^{\frac{1}{2}} & -\frac{1}{3} e^{\frac{1}{4}} & \frac{5}{6} & -\frac{1}{3} e^{\frac{1}{4}} & -\frac{1}{12} e^{\frac{1}{2}} \\
-\frac{1}{12} e^{\frac{3}{4}} & -\frac{1}{3} e^{\frac{1}{2}} & -\frac{1}{6} e^{\frac{1}{4}} & \frac{2}{3} & -\frac{1}{12} e^{\frac{1}{4}} \\
-\frac{1}{12} e^{1} & -\frac{1}{3} e^{\frac{3}{4}} & -\frac{1}{6} e^{\frac{1}{2}} & -\frac{1}{3} e^{\frac{1}{4}} & \frac{11}{12}
\end{array}\right]\left[\begin{array}{c}
u\left(x_{0}\right) \\
u\left(x_{1}\right) \\
u\left(x_{2}\right) \\
u\left(x_{3}\right) \\
u\left(x_{4}\right)
\end{array}\right]=\left[\begin{array}{c}
0 \\
\frac{1}{16} \\
\frac{1}{4} \\
\frac{9}{16} \\
1
\end{array}\right]
$$

Solving the system we find the approximate values of $u\left(x_{0}\right), u\left(x_{1}\right), u\left(x_{2}\right), u\left(x_{3}\right), u\left(x_{4}\right)$ as $u(0)=-1.2343, u(0.25)=-0.9507, u(0.5)=-0.7659, u(0.75)=-0.5845, u(1)=$ -0.4485 .

As a second example, we illustrate the method for a larger number of nodal points.

## Example 3.2:

Let $a=-1, b=1, f(x)=1, K(x, t)=x t+x^{2} t^{2}$. We implemented the method in MATLAB and present the results in Figure 1 below. The result of both the Composite Trapezoidal and Simpson rule is shown (for 10 and 20 nodal points), along with the exact solution $u(x)=1+\frac{10}{9} x^{2}$; the agreement is as expected.


Figure 1: Approximation: Number of nodal ponts: 10 (top) and 20 (bottom).
In terms of the convergence of the method, we will study the algorithm with the two composite methods used above. We know from [1] that the Composite Trapezoidal Rule (for numerical integration) converges with order $O\left(h^{2}\right)$, and the Composite Simpson's Rule with order $O\left(h^{4}\right)$, as the meshwidth $h$ tends to 0 . This carries over to the numerical solution of (3), as we illustrate in the following example.

## Example 3.3:

Let $a=0, b=\pi / 2, f(x)=\sin (x), K(x, t)=\sin (x) \cos (t)$. The exact solution is $u(x)=2 \sin (x)$, so we may calculate the errors reliably. We show in Figure 2 the convergence of the method, with the Composite Trapezoidal and Simpson Rules, as the number of nodal points $m$, increases. It is a logarithmic plot, and since we get straight lines with slope approximately $-2,-4$, respectively, we conclude that the methods converge with orders $O\left(m^{-2}\right), O\left(m^{-4}\right)$, respectively.


Figure 2: The convergence of the methods.
The fact that the error in Nyström's method depends on the order of the integration scheme as well as the smoothness of the kernel $K(x, t)$ may be seen as follows (see also [2]): subtracting (7) from (6), we have

$$
\left|u\left(x_{i}\right)-U\left(x_{i}\right)\right|=\left|\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t-\sum_{j=1}^{m} w_{j} K\left(x_{i}, x_{j}\right) U\left(x_{j}\right)\right|=O\left(m^{-p}\right),
$$

where $p$ is the order of the quadrature scheme (assuming $K(x, t)$ is sufficiently smooth). In the Composite Trapezoidal Rule, $p=2$, and in the Composite Simpson's Rule, $p=4$.

## 4 Numerical Solution of Fredholm Integro-Differential Equations of the $2^{\text {nd }}$ type

Equation (4) defines a Fredholm integro-differential equation of the $2^{\text {nd }}$ type; our goal, again, is to approximate $u$. We will consider the approximation of the solution to the problem

$$
\begin{align*}
u^{\prime}(x) & =f(x)+\int_{a}^{b} K(x, t) u(t) d t  \tag{11}\\
u(a) & =u_{a} \in \mathbb{R} .
\end{align*}
$$

We begin by constructing a grid $a=x_{1}<\ldots<x_{m+1}=b$ and substitute $x_{i}$ in equation (11):

$$
\begin{equation*}
u^{\prime}\left(x_{i}\right)=f\left(x_{i}\right)+\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t, i=1, \ldots, m+1 \tag{12}
\end{equation*}
$$

The integral in (12) is approximated by numerical quadrature, based on the nodal points $x_{1}, x_{2}, \ldots, x_{m+1}$ :

$$
\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t \approx \sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) u\left(x_{j}\right) .
$$

The derivative $u^{\prime}\left(x_{i}\right)$ is approximated using finite differences. Assuming the subdivision is uniform with meshwidth

$$
h=x_{i+1}-x_{i}, \forall i=1, \ldots, m,
$$

we use the centered difference

$$
u^{\prime}\left(x_{i}\right) \approx \frac{u\left(x_{i+1}\right)-u\left(x_{i-1}\right)}{2 h}, i=2, \ldots, m,
$$

for the internal nodal points, and for the endpoints we use

$$
u^{\prime}\left(x_{1}\right) \approx \frac{u\left(x_{2}\right)-u\left(x_{1}\right)}{h}, u^{\prime}\left(x_{m+1}\right) \approx \frac{u\left(x_{m+1}\right)-u\left(x_{m}\right)}{h},
$$

which are the forward and backward differences, respectively.
Denoting $U_{i} \approx u\left(x_{i}\right)$, we substitute the above in (12) and obtain

$$
\begin{aligned}
\frac{U_{2}-U_{1}}{h} & =f\left(x_{1}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{1}, x_{j}\right) U_{j}, \\
\frac{U_{i+1}-U_{i-1}}{2 h} & =f\left(x_{i}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) U_{j}, i=2, \ldots, m, \\
\frac{U_{m+1}-U_{m}}{h} & =f\left(x_{m+1}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{m+1}, x_{j}\right) U_{j} .
\end{aligned}
$$

In matrix form, the above system may be written as

$$
A \vec{U}=\vec{F}
$$

where $\vec{F}=\left[h f\left(x_{1}\right), 2 h f\left(x_{2}\right), \ldots, 2 h f\left(x_{m}\right), h f\left(x_{m+1}\right)\right]^{T}, \vec{U}=\left[U_{1}, \ldots, U_{m+1}\right]^{T}$, and the coefficient matrix $A$ is given by

$$
A=J-M,
$$

with

$$
J=\left[\begin{array}{ccccc}
-1 & 1 & & & \\
-1 & 0 & 1 & & \\
& \ddots & \ddots & \ddots & \\
& & -1 & 0 & 1 \\
& & & -1 & 1
\end{array}\right], M_{i, j}=\left\{\begin{array}{c}
-h w_{j} K\left(x_{1}, x_{j}\right), i=1, j=1, \ldots, m+1 \\
-2 h w_{j} K\left(x_{i}, x_{j}\right), i=2, \ldots, m, j=1, \ldots, m+1 \\
-h w_{j} K\left(x_{m+1}, x_{j}\right), i=m+1, j=1, \ldots, m+1
\end{array}\right.
$$

We note that the value of $U_{1}$ is already known from the initial condition ( $U_{1}=u(a)=u_{a}$ ), so the first row and column of $A$, as well as the first element in $\vec{F}$, are deleted before solving the linear system, yielding a vector $\vec{U} \in \mathbb{R}^{m}$ (as opposed to $\mathbb{R}^{m+1}$ ); the approximate values for $u\left(x_{i}\right), i=1, \ldots, m+1$ are given by $\left[u_{a}, U_{2}, \ldots, U_{m+1}\right]^{T} \in \mathbb{R}^{m+1}$.

## Example 4.1:

We illustrate the method for the following problem:

$$
\begin{aligned}
u^{\prime}(x) & =x e^{x}+e^{x}-x+\int_{0}^{1} x u(t) d t \\
u(0) & =0
\end{aligned}
$$

with exact solution $u(x)=x e^{x}$. We use the composite Trapezoidal Rule, and in Figure 4 we show the exact and approximate solutions, showing excellent agreement. In Figure 4, we show the convergence as the number of nodal points is increased, in a log-log scale. The slope of the resulting line gives the (expected) convergence rate of -2 .


Figure 3: Exact and approximate solutions for Example 4.1.

### 4.1 Singularly perturbed Fredholm Integro-Differential Equations of the $2^{\text {nd }}$ type

Suppose now that there is a parameter $\varepsilon \in(0,1]$ multiplying the derivative in (11), and we allow $\varepsilon \rightarrow 0$. The problem becomes

$$
\begin{align*}
\varepsilon u^{\prime}(x) & =f(x)+\int_{a}^{b} K(x, t) u(t) d t  \tag{13}\\
u(a) & =u_{a} \in \mathbb{R}
\end{align*}
$$

We begin, as usual, by constructing a grid $a=x_{1}<\ldots<x_{m+1}=b$ and substitute $x_{i}$ in equation (13):

$$
\begin{equation*}
\varepsilon u^{\prime}\left(x_{i}\right)=f\left(x_{i}\right)+\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t, i=1, \ldots, m+1 . \tag{14}
\end{equation*}
$$



Figure 4: The convergence of the method for Example 4.1.

The approximation of the integral in (14) is

$$
\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t \approx \sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) u\left(x_{j}\right),
$$

and the derivative $u^{\prime}\left(x_{i}\right)$ is approximated using finite differences. Assuming the subdivision is uniform with meshwidth $h$, we use centered differences for the internal nodal points, and for the endpoints we use forward and backward differences, as was done before.

Denoting $U_{i} \approx u\left(x_{i}\right)$, we substitute the above in (14) and obtain

$$
\begin{aligned}
\varepsilon \frac{U_{2}-U_{1}}{h} & =f\left(x_{1}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{1}, x_{j}\right) U_{j}, \\
\varepsilon \frac{U_{i+1}-U_{i-1}}{2 h} & =f\left(x_{i}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) U_{j}, i=2, \ldots, m, \\
\varepsilon \frac{U_{m+1}-U_{m}}{h} & =f\left(x_{m+1}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{m+1}, x_{j}\right) U_{j} .
\end{aligned}
$$

In matrix form, the above system may be written as

$$
A \vec{U}=\vec{F},
$$

where $\vec{F}=\left[h f\left(x_{1}\right), 2 h f\left(x_{2}\right), \ldots, 2 h f\left(x_{m}\right), h f\left(x_{m+1}\right)\right]^{T}, \vec{U}=\left[U_{1}, \ldots, U_{m+1}\right]^{T}$, and the coefficient matrix $A$ is given by

$$
A=\tilde{J}-M,
$$

with
$\tilde{J}=\left[\begin{array}{ccccc}-\varepsilon & \varepsilon & & & \\ -\varepsilon & 0 & \varepsilon & & \\ & \ddots & \ddots & \ddots & \\ & & -\varepsilon & 0 & \varepsilon \\ & & & -\varepsilon & \varepsilon\end{array}\right], M_{i, j}=\left\{\begin{array}{c}-h w_{j} K\left(x_{1}, x_{j}\right), i=1, j=1, \ldots, m+1 \\ -2 h w_{j} K\left(x_{i}, x_{j}\right), i=2, \ldots, m, j=1, \ldots, m+1 \\ -h w_{j} K\left(x_{m+1}, x_{j}\right), i=m+1, j=1, \ldots, m+1\end{array}\right.$
As before, the value of $U_{1}$ is already known from the initial condition $\left(U_{1}=\right.$ $u(a)=u_{a}$ ), so the first row and column of $A$, as well as the first element in $\vec{F}$, are deleted before solving the linear system; the approximate values for $u\left(x_{i}\right), i=$ $1, \ldots, m+1$ are given by $\left[u_{a}, U_{2}, \ldots, U_{m+1}\right]^{T} \in \mathbb{R}^{m+1}$.

We illustrate the numerical difficulties that arise when $\varepsilon \rightarrow 0$, in the following example.

## Example 4.2:

Consider the following problem:

$$
\begin{aligned}
\varepsilon u^{\prime}(x) & =f(x)+\int_{0}^{1} x t u(t) d t \\
u(0) & =0,
\end{aligned}
$$

with $f$ chosen so that the exact solution is $u(x)=1-e^{-x / \varepsilon}$. We use the Composite Trapezoidal Rule for the approximation, and in Figure 5 we show the results. For large values of $\varepsilon$, the method performs well; however, as $\varepsilon \rightarrow 0$, the approximations become inaccurate.


Figure 5: The exact and approximate solutions for Example 4.2.

Clearly, the method fails to provide an accurate approximation as $\varepsilon \rightarrow 0$, unless the stepsize $h$ is $O(\varepsilon)$. A uniform mesh is not appropriate for singularly perturbed problems.

This deterioration is due to the presence of boundary layers in $u$, which are rapidly varying solution components, in the vicinity of the boundary (or initial point in our case). We assume that the solution $u$ of (11) may be decomposed as

$$
\begin{equation*}
u=u_{S}+u_{B L}+u_{R}, \tag{15}
\end{equation*}
$$

where the smooth part $u_{S}$ is as smooth as the input data, the boundary layer part $u_{B L}$ behaves like the function $e^{-t / \varepsilon}$, and the remainder $u_{R}$ is negligible. This assumption is based on the analogous decomposition for singularly perturbed problems [8].

One non-uniform, layer adapted mesh is the so-called Shishkin mesh, which is a piecewise unform mesh commonly used in the literature for singularly perturbed problems [8]. To define the Shishkin mesh, we let

$$
\begin{equation*}
\tau=\min \{(b-a) / 2,2 \varepsilon \ln 2 m\} \tag{16}
\end{equation*}
$$

where $2 m$ is the number of subintervals we wish to use in the interval $[a, b]$. Then, $[a, b]$ is subdivided into $[a, a+\tau],[a+\tau, b]$, and in each subinterval, a uniform partition (of $m$ subintervals) is used. We set $h=\tau / m, H=(b-a-\tau) / m$ as the two different meshwidths. (See Figure 6 below.)

Collocating at $x_{i}, i=1, \ldots, 2 m+1$, gives

$$
\begin{equation*}
\varepsilon u^{\prime}\left(x_{i}\right)=f\left(x_{i}\right)+\int_{a}^{b} K\left(x_{i}, t\right) u(t) d t, i=1, \ldots, 2 m+1 . \tag{17}
\end{equation*}
$$



Figure 6: The Shishkin mesh.

For the approximation of the integral in (17) we have

$$
\int_{a}^{b} K(x, t) u(t) d t=\int_{a}^{a+\tau} K(x, t) u(t) d t+\int_{a+\tau}^{b} K(x, t) u(t) d t,
$$

with

$$
\begin{gathered}
\int_{a}^{a+\tau} K\left(x_{i}, t\right) u(t) d t \approx \sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) u\left(x_{j}\right), \\
\int_{a}^{a+\tau} K\left(x_{i}, t\right) u(t) d t \approx \sum_{j=1}^{m+1} W_{j} K\left(x_{i}, x_{j+m}\right) u\left(x_{m+j}\right),
\end{gathered}
$$

where $w_{j}, W_{j}$ are the weights in the chosen composite quadrature rule on $[a, a+$ $\tau],[a+\tau, b]$, respectively. The discrete equations become

$$
\begin{equation*}
\varepsilon u_{i}^{\prime}=f\left(x_{i}\right)+\sum_{j=1}^{m+1} w_{j} K\left(x_{i}, x_{j}\right) u_{j}+\sum_{j=1}^{m+1} W_{j} K\left(x_{i}, x_{j+m}\right) u_{m+j} . \tag{18}
\end{equation*}
$$

The derivative $u_{i}^{\prime}$ above is approximated using finite differences as before forward difference for the first point, backward difference for the last point, and centered difference for the internal points. However, for $u^{\prime}\left(x_{m+1}\right)$ the meshwidth is different in the two intervals surrounding $x_{m+1}=a+\tau$ (see Figure 6), hence the centered difference for this point is modified as follows [10]:

$$
u_{m+1}^{\prime} \approx \frac{u_{m+2}-(H / h)^{2} u_{m}-\left(1-(H / h)^{2}\right) u_{m+1}}{H(1+H / h)} .
$$

The resulting system, in matrix form, may be written as

$$
A \vec{U}=\vec{F},
$$

where $\vec{F}=\left[h f\left(x_{1}\right), 2 h f\left(x_{2}\right), \ldots, 2 H f\left(x_{2 m}\right), H f\left(x_{2 m+1}\right)\right]^{T}, \vec{U}=\left[U_{1}, \ldots, U_{2 m+1}\right]^{T}$, $\mathrm{j}=1, \ldots, 2 \mathrm{~m}+1$, and the coefficient matrix $A$ is given by

$$
A=\left[\begin{array}{ccc}
J_{1}-M_{1} & \vec{a}_{1} & -M_{2} \\
\vec{a}_{2} & \alpha & \vec{a}_{3} \\
-M_{3} & \vec{a}_{4} & J_{2}-M_{4}
\end{array}\right],
$$

where

$$
\begin{gathered}
J_{1}=\left[\begin{array}{ccccc}
-\varepsilon & \varepsilon & & & \\
-\varepsilon & 0 & \varepsilon & & \\
& \ddots & \ddots & \ddots & \\
& & -\varepsilon & 0 & \varepsilon \\
& & & -\varepsilon & 0
\end{array}\right] \in \mathbb{R}^{m \times m}, J_{2}=\left[\begin{array}{cccc}
0 & \varepsilon & & \\
-\varepsilon & 0 & \varepsilon & \\
& \ddots & \ddots & \ddots \\
& & -\varepsilon & 0
\end{array}\right] \\
\\
\\
\\
{\left[M_{1}\right]_{i, j}=} \\
\\
{\left[M_{2}\right]_{i, j}=\left\{\begin{array}{c}
h w_{j} K\left(x_{1}, x_{j}\right), i=1, j=1, \ldots, m \\
2 h w_{j} K\left(x_{i}, x_{j}\right), i=2, \ldots, m, j=1, \ldots, m
\end{array}\right.} \\
{\left[\begin{array}{c}
h W_{j+1} K\left(x_{1}, x_{j+m+1}\right), i=1, j=1, \ldots, m \\
2 h W_{j+1} K\left(x_{i}, x_{j+m+1}\right), i=2, \ldots, m, j=1, \ldots, m
\end{array}\right.} \\
{\left[M_{3}\right]_{i, j}=\left\{\begin{array}{c}
2 H w_{j} K\left(x_{m+1+i}, x_{j}\right), i=1, j=1, \ldots, m \\
H w_{j} K\left(x_{2 m+1}, x_{j}\right), i=m, j=1, \ldots, m
\end{array}\right.} \\
{\left[M_{4}\right]_{i, j}=\left\{\begin{array}{c}
2 H W_{j+1} K\left(x_{m+1+i}, x_{m+1+j}\right), i=1, \ldots, m-1 \\
, j=1, \ldots, m \\
H W_{j+1} K\left(x_{2 m+1}, x_{m+1+j}\right), j=1, \ldots, m
\end{array}\right.}
\end{gathered}
$$

and

$$
\vec{a}_{1}=\left[c_{1}, \vec{V}_{1}, c_{2}\right]^{T}, \vec{a}_{2}=\left[\vec{V}_{2}, c_{3}\right], \vec{a}_{3}=\left[c_{4}, \vec{V}_{3}\right], \vec{a}_{4}=\left[c_{5}, \vec{V}_{4}, c_{6}\right]^{T}
$$

$$
\begin{gathered}
{\left[\vec{V}_{1}\right]_{k}=-2 h\left(w_{m+1}+W_{1}\right) K\left(x_{k+1}, x_{m+1}\right), k=1, \ldots, m-2} \\
{\left[\vec{V}_{4}\right]_{k}=-2 H\left(w_{m+1}+W_{1}\right) K\left(x_{k+2+m}, x_{m+1}\right), k=1, \ldots, m-2} \\
{\left[\vec{V}_{2}\right]_{j}=-H(1+H / h) w_{j} K\left(x_{m+1}, x_{j}\right), j=1, \ldots, m-1} \\
{\left[\vec{V}_{3}\right]_{j}=-H(1+H / h) W_{j+2} K\left(x_{m+1}, x_{j+m+2}\right), j=1, \ldots, m-1}
\end{gathered}
$$

$$
\begin{gathered}
c_{1}=-h\left(w_{m+1}+W_{1}\right) K\left(x_{1}, x_{m+1}\right) \\
c_{2}=\varepsilon-2 h\left(w_{m+1}+W_{1}\right) K\left(x_{m}, x_{m+1}\right) \\
c_{3}=-\varepsilon(H / h)^{2}-H\left(1+(H / h) w_{m}+W_{1}\right) K\left(x_{m+1}, x_{m}\right) \\
c_{4}=\varepsilon-H\left(1+(H / h) W_{2} K\left(x_{m+1}, x_{m+2}\right)\right.
\end{gathered}
$$

$$
\begin{gathered}
c_{5}=-\varepsilon-2 H\left(w_{m+1}+W_{1}\right) K\left(x_{m+2}, x_{m+1}\right) \\
c_{6}=-H\left(w_{m+1}+W_{1}\right) K\left(x_{2 m+1}, x_{m+1}\right) \\
\alpha=-\varepsilon\left(1-(H / h)^{2}\right)-H\left(1+(H / h) W_{m+1} K\left(x_{m+1}, x_{m+1}\right)\right.
\end{gathered}
$$

We re-visit the problem in Example 4.2 in the following.

## Example 4.3

Consider the problem:

$$
\begin{aligned}
\varepsilon u^{\prime}(x) & =f(x)+\int_{0}^{1} x t u(t) d t \\
u(0) & =0,
\end{aligned}
$$

with $f$ chosen so that the exact solution is $u(x)=1-e^{-x / \varepsilon}$. We use the Composite Trapezoidal Rule for the approximation, on the Shishkin mesh, as described above. In Figure 7 we show the results for $\varepsilon=0.025,0.01$, and no deterioration is observed (as was the case in Example 4.2).


Figure 7: The exact and approximate solutions for Example 4.3.
We also show the convergence of the method for various values of $\varepsilon$, in Figure 8. We first observe that the method is robust (with respect to $\varepsilon$ ). Next, the slope calculated is slightly lower than -2 , as was observed for non-singularly perturbed problems. The reason for this is that the Shishkin mesh yields a convergence rate of order $O\left(m^{-2} \ln ^{2}(m)\right)$ as opposed to $O\left(m^{-2}\right)$, with the logarithmic term not removable (see, e.g. [8]).


Figure 8: Convergence of the method for Example 4.3.

As a final example, we consider the following.

## Example 4.4

$$
\begin{aligned}
\varepsilon u^{\prime}(x) & =f(x)+\int_{0}^{1}(\cos (x)+\sin (t)) u(t) d t \\
u(0) & =0
\end{aligned}
$$

with $f$ chosen so that the exact solution is $u(x)=\cos (\pi x)-e^{-x / \varepsilon}$. We use the Composite Trapezoidal Rule for the approximation, on the Shishkin mesh, and in Figure 9 we show the result for $\varepsilon=0.01$.


Figure 9: Exact and approximate solutions for Example 4.4.


Figure 10: Convergence for Example 4.4.

The convergence of the method for various values of $\varepsilon$, is shown in Figure 10. The observations are exactly the same as the previous example.

## Example 4.5

We finally consider a problem without a known exact solution:

$$
\begin{aligned}
\varepsilon u^{\prime}(x) & =1-e^{-x / e}+\int_{0}^{1}\left(t+e^{-x / e}\right) u(t) d t \\
u(0) & =0
\end{aligned}
$$

In Figure 11 we show the approximate solution for various $\varepsilon$, and we see that the method produces the expected results.


Figure 11: Approximate solution for Example 4.5.

## 5 Closing Remarks

In this thesis we considered Fredholm integral and integro-differential equations of the second type, and the numerical approximation of their solution using Nyström's method of collocation, in conjunction with the composite Trapezoidal and Simpson's rules - we mainly utilized the Trapezoidal Rule.

For Fredholm integral equations, we implemented the method and performed numerical experiments for problems with smooth data (hence smooth solution). The convergence rates observed, for both composite quadrature rule choices, namely $O\left(m^{-2}\right)$ and $O\left(m^{-4}\right)$, respectively, where $m$ is the number of nodal points, were the expected ones.

We also considered integro-differential equations of first order, and for the derivative approximation we used finite differences. In particular, we used centered differences for the internal nodal points, and forward and backward differences for the first and last points, respectively. For smooth data, the method was observed to yield the expected $O\left(\mathrm{~m}^{-2}\right)$ convergence rate, when the Trapezoidal Rule was used for the integral.

The above convergence rate did not hold when we considered singularly perturbed integro-differential equations. The method "broke down" due to the fact that we were using a uniform mesh to approximate a solution which includes boundary layers. This was rectified with the use of a layer-adapted, piecewise uniform Shishkin mesh [8]. We verified through numerical experiments that with this choice of the mesh points, Nyström's method yields uniform convergence, independently of $\varepsilon$, as the number of mesh points is increased, at the rate $O\left(m^{-2} \mathrm{ln}^{2} m\right)$ (for the Trapezoidal Rule).

The next step in the study of this method for such problems is to prove the observed convergence rates. Once this is done, the case of systems could be considered, with either one or different singular perturbation parameters.

One could also study Galerkin variational methods [7] for the numerical approximation of the solution to such problems. In the Galerkin method for, e.g (11), we multiply the equation with a so-called test function $v(x)$ and integrate over the interval $[a, b]$, to get

$$
\begin{equation*}
\int_{a}^{b} v(x) \varepsilon u^{\prime}(x) d x=\int_{a}^{b} v(x) f(x) d x+\int_{a}^{b} \int_{a}^{b} v(x) K(x, t) u(t) d t d x \tag{19}
\end{equation*}
$$

The variational problem to be solved is: find $u \in U$ such that (19) holds for $v \in V$, where $U, V$ are appropriate function spaces (e.g. derivatives are squared integrable functions, etc.). At the discrete level, we seek $u_{N} \in U_{N} \subset U$ such that (19) holds for all $v \in V_{N}$, where $U_{N}, V_{N}$ are appropriate finite dimensional subspaces of $U, V$, respectively. Assuming

$$
U_{N}=\operatorname{span}\left\{\phi_{1}, \ldots, \phi_{N}\right\}, V_{N}=\operatorname{span}\left\{\psi_{1}, \ldots, \psi_{N}\right\}
$$

with $\phi, \psi$ polynomials, we have

$$
u_{N}=\sum_{i=1}^{N} \xi_{i} \phi_{i}, v_{N}=\sum_{i=1}^{N} \zeta_{i} \psi_{i} .
$$

Substituting the above expressions in (19), we obtain a linear system for the coefficients $\xi_{i}, i=1, \ldots, N$, hence we obtain the approximate solution $u_{N}$. This study will appear elsewhere.

## References

[1] K. E. Atkinson, An Introduction to Numerical Analysis, John Wiley and Sons, 1989.
[2] K. E. Atkinson, The Numerical Solution of Integral Equations of the Second Kind, Cambridge University Press, New York, 1997.
[3] V. E. Barlette, M. M. Leite, and S. K. Adhikari, Integral equations of scattering in one dimension, American Journal of Physics, Vol. 69, Mo. 9, pp. 1010-1013, 2001.
[4] K. Herrebrugh, Solving the incompressible and isothermal problem in elastohydrodynamic lubrication through an integral equation, Journal of Lubrication Technology, Vol. 90, No. 1, pp. 262-270, 1968.
[5] N. Keyfitz, The integral equation of population analysis. Review of the International Statistical Institute, Vol. 35, No. 3, pp. :213-246, 1967.
[6] A. A. Men, On an integral equation of radiation-conduction heat transfer, Journal of Engineering Physics and Thermophysics, Vol. 18, No. 3, pp. 327332, 1970.
[7] S. Repin, 100 years of the Galerkin method, Comput. Methods Appl. Math., Vol. 17, No. 3, pp. 351-357, 2017.
[8] H. G. Roos, M. Stynes and L. Tobiska, Robust Numerical Methods for Singularly Perturbed Differential Equations. Convection-Diffusion-Reaction and Flow Problems, Springer Series in Computational Mathematics (SSCM, volume 24), 2008.
[9] J. L. Stephenson, Theory of transport in linear biological systems: I. Fundamental integral equation, The bulletin of mathematical biophysics, Vol 22, pp. 1-17, 1960.
[10] H. Sundqvist and G. Veronis, A simple finite-difference grid with nonconstant intervals, Tellus, Vol. 22, No. 1, pp. 26-31, 1970.

