



A New Class of Fredholm Integral Equations of the Second Kind with Non Symmetric Kernel: Solving by Wavelets Method

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ABSTRACT: In this paper, we introduce an efficient modification of the wavelets method to solve a new class of Fredholm integral equations of the second kind with non symmetric kernel. This method based on orthonormal wavelet basis, as a consequence three systems are obtained, a Toeplitz system and two systems with condition number close to 1. Since the preconditioned conjugate gradient normal equation residual (CGNR) and preconditioned conjugate gradient normal equation error (CGNE) methods are applicable, we can solve the systems in $O(2n \log(n))$ operations, by using the fast wavelet transform and the fast Fourier transform.

Key Words: Fredholm integral equation, Non symmetric kernel, Wavelet basis, Toeplitz matrix, Condition number.

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1. Introduction

Integral equation perform role effectively in many fields of science and engineering. Recently, there are a lot of orthonormal basis function that have been used to find an approximate solution, mention Fourier functions [2], Legendre polynomials

[21] and wavelets [10,12,13,16,17,19,20,26]. Although, the wavelet bases are one of the most interesting basis, especially for large scale problems, in which the kernel can be constituted as sparse matrix.

We reminder that usually it is difficult to construct the exact solution of linear and nonlinear Fredholm integral equation via the well-known methods. A lot of different useful methods have been developed to approximate the solutions of these equations. For instance, collocation methods are studied in [15,24], spectral methods are given in [14,18], transform methods are introduced in [1,3,23], and homotopy perturbation method is presented in [8] and others.

More recently, the multiresolution analysis has been considered by many researchers (see [11,12,17,19,20,28]). We mention that wavelets method play a key role to find the unique solution for some Fredholm integral equations.

In the present paper, we present wavelet basis to find the approximate solution of the following Fredholm integral equation of second kind:

$$u(t) - 2^\beta \int_0^{+\infty} k(2^\alpha s - 2^\alpha t)u(t)dt = f(t), \quad s \in [0, +\infty[, \quad \alpha > 0, \quad \beta \in \mathbb{R}, \quad (1.1)$$

where $u(\cdot)$ is the unknown function, $f(\cdot)$ is the known function and $k(s - t)$ is a non symmetric kernel.

A considerable part of this proposal is based on a study by [Jin and Yuan, 1998], in which the authors focused on new class the first kind with symmetric kernel. In contrast to their work, we focused on the second kind with non symmetric kernel and as we know that the symmetric property is necessary condition to apply conjugate gradient method and in our case we don't have this property so we dealt with the equivalent two systems that have the symmetric property.

The outline of the paper is as follows: In section 2, we describe the basic formulation of wavelets and preliminary which are necessary for our development. Section 3 is devoted to the discretization of the integral equation. In section 4, we study the condition number of the matrix operator and we give the operation cost to solve the systems.

2. Preliminaries

2.1. Wavelet bases

The basic tool for our method to approximate the solution of (1.1) is wavelet Bases. For the convenience of the reader, we recall here some basic concepts and well-known results concerning the multiresolution analysis (MRA for short). As in [7,11], let us consider a function $\varphi \in L^2(\mathbb{R})$ called the father wavelet (or scaling function), with a compact support $[0, a]$, $a > 0$. We assume that

$$\varphi(\cdot - k), \quad k \in \mathbb{Z} \quad (2.1)$$

form an orthonormal sequence in $L^2(\mathbb{R})$. Let V_0 be the closed linear subspace of $L^2(\mathbb{R})$ generated by (2.1). The multiresolution analysis (MRA), depending on the $\varphi(\cdot)$ consists of:

- (i) $f(.) \in V_0$ if and only if $f(2^j.) \in V_j$ for all $j \in \mathbb{Z}$;
- (ii) $\cdots \subset V_{j-1} \subset V_j \subset V_{j+1} \subset \cdots$;
- (iii) $\overline{\bigcup_{j \in \mathbb{Z}} V_j} = L^2(\mathbb{R})$ and $\bigcap_{j \in \mathbb{Z}} V_j = \{0\}$;

(iv) The sequence (2.1) forms a Riesz basis of V_0 .

Let W_j be the orthogonal complement of V_j in V_{j+1} , i.e.,

$$V_{j+1} = V_j \oplus W_j.$$

According to the above definition, we have

$$\bigoplus_{-\infty}^{+\infty} W_j = L^2(\mathbb{R}).$$

Following [6,11,22], there exists at least one function $\psi \in W_0$ such that

$$\psi(\cdot - k), \quad k \in \mathbb{Z}$$

is an orthonormal basis of W_0 . The function ψ is called the mother wavelet.

A wavelet $\phi \in L^2(\mathbb{R})$ is called orthonormal if the family of functions generated from ϕ by

$$\phi_{j,k}(s) = 2^{j/2} \phi(2^j s - k), \quad j, k \in \mathbb{Z},$$

is orthonormal, that is,

$$\langle \phi_{j,k}, \phi_{m,n} \rangle = \delta_{j,m} \delta_{k,n}.$$

Let us introduce the following two wavelet sequences:

$$\varphi_{j,k}(s) = 2^{j/2} \varphi(2^j s - k), \quad j, k \in \mathbb{Z},$$

and

$$\psi_{j,k}(s) = 2^{j/2} \psi(2^j s - k), \quad j, k \in \mathbb{Z}.$$

We recall that

$$\langle \psi_{m,k}, \varphi_{m,l} \rangle = \langle \psi_{n,k}, \varphi_{n,l} \rangle, \quad \text{for all } m, n, k, l \in \mathbb{Z}.$$

Therefore, the wavelet sequence $\{\psi_{j,k}\}$ forms a Riesz basis of $H^s(\mathbb{R})$ for $s \geq 0$.

Assume that \mathbb{B}_1 and \mathbb{B}_2 two bases in V_n with:

$$\mathbb{B}_1 = (\varphi_{n,k}(\cdot))_k, \quad k \in \mathbb{Z},$$

and

$$\mathbb{B}_2 = \bigcup_{-\infty < j \leq n-1} (\psi_{j,k}(\cdot))_k, \quad k \in \mathbb{Z}.$$

We note that \mathbb{B}_1 and \mathbb{B}_2 follow from the father wavelet φ and the mother wavelet ψ , respectively.

2.2. Wavelet transform

Definition 2.1 (Continuous wavelet transform). *The continuous wavelet transform of the mother wavelet φ is defined by*

$$(S_\varphi f)(j, k) = \int_{-\infty}^{+\infty} f(t) \overline{\varphi_{j,k}(t)} dt = \langle f, \varphi_{j,k} \rangle.$$

Definition 2.2 (Discrete wavelet transform). *The discrete wavelet transform of the father wavelet ψ is defined by*

$$(S_\psi f)(j, k) = \int_{-\infty}^{+\infty} f(t) \overline{\psi_{j,k}(t)} dt = \langle f, \psi_{j,k} \rangle.$$

2.3. Condition number

Condition number of a matrix gives the information about the singularity of the corresponding matrix.

Definition 2.3 (Condition number). *Let A be an $n \times n$ invertible matrix. Define $\kappa(A)$, the condition number of A , by*

$$\kappa(A) = \|A\| \|A^{-1}\|.$$

The condition number of an $n \times n$ invertible matrix A is defined as the ratio of its maximum singular value to its minimum singular value, that is, for

$$\lambda_M := \max \{ |\lambda|, \lambda \text{ is an eigenvalue of } A \},$$

and

$$\lambda_m := \min \{ |\lambda|, \lambda \text{ is an eigenvalue of } A \},$$

we have

$$\kappa(A) = \frac{\lambda_M}{\lambda_m}.$$

2.4. Preconditioning and diagonal scaling

A preconditioner P of a matrix A is given by $P^{-1}A$ which its condition number smaller than the original matrix. In order to solve linear systems of the form $Ax = b$, preconditioners are used for numerous iterative methods. Then, while the condition number of the matrix A decreases, for a lot of iterative linear system solvers the rate of convergence increases.

Hence, preconditioning is a very effective tool which uses to reduce the condition number of the matrix A .

Diagonal scaling (DS) is a special case of preconditioning and it is an efficient tool used to reduce the condition number of matrix A for ensuring the convergence and the accuracy of the first method. In our case, to reduce the condition number of the matrix A we apply the diagonal matrix D , in a way to speed up the method.

2.5. Conjugate gradient method

Conjugate gradient (CG) method uses to solve linear system of the form $Ax = b$, this method can be used also to obtain a quick convergence when $\kappa(A)$ is smaller.

Generally, conjugate gradient method uses for solving large problems in order to attain a modest accuracy in a reasonable number of iterations.

2.5.1. Conjugate gradient normal equation residual and error. The conjugate gradient method can be applied to solve the normal equations. The CGNE and CGNR methods are important variants of this approach, which are the simplest methods for non symmetric or indefinite systems. Since other methods for such systems are in general rather more complicated than the conjugate gradient method. These methods transform a linear system to a symmetric definite one for applying the conjugate gradient method.

CGNR solves the system

$$(A^T A)x = A^T b.$$

CGNE solves the system

$$(AA^T)y = d.$$

3. Discretization of integral equation

Let $H^s(\mathbb{R})$ and $H^t(\mathbb{R})$ be two Sobolev spaces, with $s \geq t \geq 0$. Letting

$$(Ku)(s) := 2^\beta \int_0^{+\infty} k(2^\alpha s - 2^\alpha t)u(t)dt, \quad (3.1)$$

we assume that $k(2^\alpha \cdot - 2^\alpha \cdot) \in H^s(\mathbb{R})$ is continuous non symmetric kernel.

The integral operator K from $H^s(\mathbb{R})$ into $H^t(\mathbb{R})$ is compact.

Eq. (1.1) can be rewritten in operator form as follows:

$$(I - K)u = f.$$

We assume that 1 is not a spectrum value of K . Hence, the equivalent variational form follows:

$$\begin{cases} \text{find } u \in H^s(\mathbb{R}), & \text{such that} \\ B(u, v) = F(v), & v \in H^s(\mathbb{R}), \end{cases} \quad (3.2)$$

where

$$\begin{aligned} B(u, v) &:= \langle u, v \rangle - \langle Ku, v \rangle \\ &= \int_0^{+\infty} u(s)v(s)ds - \int_0^{+\infty} \int_0^{+\infty} k(s-t)u(t)v(s)dsdt, \end{aligned}$$

and

$$F(v) := \int_0^{+\infty} f(s)v(s)ds.$$

Since

$$\langle Ku, v \rangle \leq \beta \|Ku\|_{H^t} \|v\|_{H^t},$$

it follows that $\langle Ku, v \rangle$ is a continuous bilinear form on $H^t(\mathbb{R}) \times H^s(\mathbb{R})$.
We assume that

$$\langle Ku, v \rangle \geq \rho \|u\|_{H^s}^2, \quad \text{for some constant } \rho > 0.$$

Hence, $\langle Ku, v \rangle$ is coercive form on $H^t(\mathbb{R}) \times H^s(\mathbb{R})$.

3.1. Projection of $(I - A)$ with respect to \mathbb{B}_1 and \mathbb{B}_2

- Let the matrix $(I - A_n)$ relative to the basis \mathbb{B}_1 , which is the projection of the matrix $(I - A)$ on the subspace V_n .

The elements of the matrix $(I - A_n)$ are given as follows

$$\begin{aligned} t_{p,q} &:= \langle \varphi_{n,p}, \varphi_{n,q} \rangle - \langle K\varphi_{n,p}, \varphi_{n,q} \rangle \\ &= \int_0^{+\infty} \varphi_{n,p}(s) \varphi_{n,q}(s) ds - 2^\beta \int_0^{+\infty} \int_0^{+\infty} k(2^\alpha s - 2^\alpha t) \varphi_{n,p}(t) \varphi_{n,q}(s) dt ds. \end{aligned} \quad (3.3)$$

For all $u, v \in H^s(\mathbb{R})$, we assume that u_n, v_n are the projections of u, v on V_n respectively. Which implies that (3.2) becomes

$$\int_0^{+\infty} u_n(s) v_n(s) ds - \int_0^{+\infty} \int_0^{+\infty} k(s - t) u_n(t) v_n(s) dt ds = \int_0^{+\infty} f(s) v_n(s) ds. \quad (3.4)$$

Let

$$u_n = \sum_p x_p \varphi_{n,p} \quad \text{and} \quad v_n = \varphi_{n,q}, \quad \text{for all } q \in \mathbb{Z}. \quad (3.5)$$

By substituting (3.5) into (3.4), we get a linear system given as follows

$$(I - T_\infty)x = b, \quad (3.6)$$

where $(I - T_\infty)_{p,q} = t_{p,q}$ is given by (3.3), and

$$(x)_p = x_p, \quad (b)_q = \int_0^{+\infty} f(s) \varphi_{n,q}(s) ds.$$

We mention that φ has the compact support $[0, a]$, which leads us to $t_{p,q} = t_{p-q}$.

$$\begin{aligned}
t_{p,q} &= \int_0^{+\infty} \varphi_{n,p}(s) \varphi_{n,q}(s) ds - \int_0^{+\infty} \int_0^{+\infty} 2^\beta k(2^\alpha s - 2^\alpha t) \varphi_{n,p}(t) \varphi_{n,q}(s) dt ds \\
&= \delta_{p,q} - 2^{\beta+n} \int_0^{+\infty} \int_0^{+\infty} k(2^\alpha s - 2^\alpha t) \varphi(2^n t - p) \varphi(2^n s - q) dt ds \\
&= \delta_{p,q} - 2^{\beta+n} \int_{2^{-n}p}^{2^{-n}(a+p)} \int_{2^{-n}q}^{2^{-n}(a+q)} k(2^\alpha s - 2^\alpha t) \varphi(2^n t - p) \varphi(2^n s - q) dt ds \\
&= \delta_{p,q} - 2^\beta \times 2^{-n} \int_0^a \int_0^a k[2^{-n} \times 2^\alpha (s - t + p - q)] \varphi(t) \varphi(s) dt ds \\
&= \delta_{p,q} - 2^{-n+\beta} \int_0^a \int_0^a k[2^{-n+\alpha} (s - t + p - q)] \varphi(t) \varphi(s) dt ds \\
&= t_{p-q}.
\end{aligned}$$

Hence $(I - T_\infty)$ is a Toeplitz matrix.

• The matrix representation of $(I - A_n)$ relative to the basis \mathbb{B}_2 has the elements given as follows

$$\begin{aligned}
a_{p,q,i,j} &:= \langle \psi_{p,q} \psi_{i,j} \rangle - \langle K \psi_{p,q}, \psi_{i,j} \rangle \\
&= \int_0^{+\infty} \psi_{p,q}(s) \psi_{i,j}(s) ds \\
&\quad - 2^\beta \int_0^{+\infty} \int_0^{+\infty} k(2^\alpha s - 2^\alpha t) \psi_{p,q}(t) \psi_{i,j}(s) dt ds,
\end{aligned} \tag{3.7}$$

for $-\infty < p, i < n$ and $-\infty < q, j < +\infty$.

Writing

$$u_n = \sum_{p,q} x_{p,q} \psi_{p,q}, \quad \text{and} \quad v_n = \psi_{p,q}, \quad -\infty < p < n, \quad \text{for all } q \in \mathbb{Z}. \tag{3.8}$$

We substitute (3.8) into (3.4), we obtain the linear system

$$(I - A_\infty)x = d, \tag{3.9}$$

where $(I - A_\infty)_{p,q,i,j} = a_{p,q,i,j}$ is unsymmetric given by (3.7), $x = (x_{p,q})^T$ and $d = (d_{p,q})^T$ are vectors with $d_{p,q} := \int_0^{+\infty} f(s) \psi_{p,q}(s) ds$.

4. Solving the linear systems

4.1. Condition number

From the previous section we obtained two different linear systems. One of them is the Toeplitz system (3.6) (relative to \mathbb{B}_1) and the other one is the systems

(3.9) (relative to \mathbb{B}_2).

Let us focus on studying the condition number of the last linear system. Actually, we will develop the idea of Zhang [28]. In order to do that, firstly, we present the following Lemma which plays an important role for reducing the condition number of the matrix.

Lemma 4.1. ([11, 22, 28]) *Let*

$$f = \sum_{j,k} \langle f, \psi_{j,k} \rangle \psi_{j,k}.$$

Then $f \in H^s(\mathbb{R})$ if and only if

$$\sum_{j,k} |\langle f, \psi_{j,k} \rangle|^2 (1 + 4^{js}) < +\infty, \quad -r < s < r,$$

where r is the regularity of the MRA. Moreover, since $\{\psi_{j,k}\}$ is a Riesz basis of $H^s(\mathbb{R})$, we also have

$$C_1 \sum_{j,k} |\langle f, \psi_{j,k} \rangle|^2 (1 + 4^{js}) \leq \|f\|_{H^s}^2 \leq C_2 \sum_{j,k} |\langle f, \psi_{j,k} \rangle|^2 (1 + 4^{js}), \quad (4.1)$$

where $C_2 \geq C_1 > 0$ are constants.

Secondly, we know that $(I - A_\infty)$ in system (3.9) is unsymmetric. Then, system (3.9) becomes

$$(I - A_\infty)^T (I - A_\infty) x = (I - A_\infty)^T d, \quad (4.2)$$

$$(I - A_\infty)(I - A_\infty)^T y = d, \quad x = (I - A_\infty)^T y. \quad (4.3)$$

Now, let $\phi \in V_n$ with $\phi = \sum_{j,k} w_{j,k} \psi_{j,k}$. We have

$$\begin{aligned} B_1(\phi, \phi) &:= \sum_{j,k} \sum_{i,\ell} w_{j,k} w_{i,\ell} [\langle (I - A_\infty)^T (I - A_\infty) \psi_{j,k}, \psi_{i,\ell} \rangle] \\ &= w^T (I - A_\infty)^T (I - A_\infty) w, \end{aligned} \quad (4.4)$$

and

$$\begin{aligned} B_2(\phi, \phi) &:= \sum_{j,k} \sum_{i,\ell} w_{j,k} w_{i,\ell} [\langle (I - A_\infty)(I - A_\infty)^T \psi_{j,k}, \psi_{i,\ell} \rangle] \\ &= w^T (I - A_\infty)(I - A_\infty)^T w, \end{aligned} \quad (4.5)$$

where $w := (w_{j,k})^T$ is a vector. By the assumption that

$$B(u, v) \in \{B_1(u, v), B_2(u, v)\}$$

is a continuous elliptic bilinear form on the space $H^s(\mathbb{R}) \times H^s(\mathbb{R})$, i.e.,

$$\begin{aligned} B(u, v) &\leq \beta \|u\|_{H^s} \cdot \|v\|_{H^s}, \\ B(u, u) &\geq \alpha \|u\|_{H^s}^2. \end{aligned}$$

Since $\phi \in V_j$, we get $\phi \in H^s$.

Consequently,

$$C_3 \|\phi\|_{H^s}^2 \leq B(\phi, \phi) \leq C_4 \|\phi\|_{H^s}^2, \quad \text{for some constants } C_4 \geq C_3 > 0. \quad (4.6)$$

4.1.1. *Condition number of system (4.4).* From (4.4) and (4.6), we get

$$C_3 \|\phi\|_{H^s}^2 \leq w^T (I - A_\infty)^T (I - A_\infty) w \leq C_4 \|\phi\|_{H^s}^2.$$

By using (4.1), we obtain

$$C_1 \sum_{j,k} |\langle w, \psi_{j,k} \rangle|^2 (1 + 4^{js}) \leq \|\phi\|_{H^s}^2 \leq C_2 \sum_{j,k} |\langle w, \psi_{j,k} \rangle|^2 (1 + 4^{js}),$$

then

$$C_1 \sum_{j,k} |w_{j,k}|^2 2^{2js} \leq \|\phi\|_{H^s}^2 \leq C_2 \sum_{j,k} |w_{j,k}|^2 + C_2 \sum_{j,k} |w_{j,k}|^2 2^{2js}.$$

Thus,

$$C_1 \sum_{j,k} |2^{js} w_{j,k}|^2 \leq \|\phi\|_{H^s}^2 \leq C_0 \sum_{j,k} |2^{js} w_{j,k}|^2,$$

so that

$$\begin{aligned} C_3 C_1 \sum_{j,k} |2^{js} w_{j,k}|^2 &\leq C_3 \|\phi\|_{H^s}^2 \leq w^T (I - A_\infty)^T (I - A_\infty) w \\ &\leq C_4 \|\phi\|_{H^s}^2 \leq C_4 C_0 \sum_{j,k} |2^{js} w_{j,k}|^2. \end{aligned}$$

Consequently,

$$C_5 \sum_{j,k} |2^{js} w_{j,k}|^2 \leq w^T (I - A_\infty)^T (I - A_\infty) w \leq C_6 \sum_{j,k} |2^{js} w_{j,k}|^2,$$

for some constants $C_5 \geq C_6 > 0$.

By using diagonal scaling D , we get

$$C_5 \|w\|^2 \leq w^T D^{-1/2} (I - A_\infty)^T (I - A_\infty) D^{-1/2} w \leq C_6 \|w\|^2,$$

where $\|\cdot\|$ is the L^2 -norm. In the end, the condition number of $(I - A_\infty)^T (I - A_\infty)$ is close to 1, that is,

$$k(D^{-1/2} (I - A_\infty)^T (I - A_\infty) D^{-1/2}) = O(1).$$

4.1.2. *Condition number of system (4.5).* From (4.5) and (4.6), we get

$$C_3 \|\phi\|_{H^s}^2 \leq w^T (I - A_\infty)(I - A_\infty)^T w \leq C_4 \|\phi\|_{H^s}^2.$$

By following the same steps of the previous system we obtain that the condition number of $(I - A_\infty)(I - A_\infty)^T$ after a diagonal scaling is

$$k(D^{-1/2}(I - A_\infty)(I - A_\infty)^T D^{-1/2}) = O(1).$$

4.2. Operation cost of the corresponding systems

In order to numerically solve the system (3.6), we use a finite interval. For this reason, let us consider the finite section T_n of T_∞ . Thus, the Toeplitz system (3.6) becomes an $n - by - n$ system

$$(I - T_n)x = b. \quad (4.7)$$

Now, we introduce the relation between $(I - T_n)$ and $(I - A_n)$, which is similar to the one given by the authors of [11] as follows

$$(I - A_n) = W_n(I - T_n)W_n^{-1},$$

where $(I - A_n)$ is the finite section of $(I - A_\infty)$ and W_n is a finite section of W which is the wavelet transform matrix between two orthonormal wavelet bases \mathbb{B}_1 and \mathbb{B}_2 .

Hence, we solve the Toeplitz system (3.6) by solving its equivalent form

$$(W_n(I - T_n)W_n^{-1}) W_n x = W_n b,$$

i.e.,

$$(I - A_n)\tilde{x} = \tilde{b}, \quad (4.8)$$

where $\tilde{x} := W_n x$ and $\tilde{b} := W_n b$.

Now, we are going to solve the system (4.8). However, the matrix $(I - A_n)$ does not have a small condition number. Then we would like to apply PCG method with diagonal preconditioner D_n in order to obtain a new matrix with a smaller condition number. Unfortunately, $(I - A_n)$ does not have the symmetric property. That means the PCG method will not work. Thus, two systems are obtained with symmetric property.

$$(I - A_n)_n^T (I - A_n) \tilde{x} = (I - A_n) T \tilde{b}, \quad (4.9)$$

$$(I - A_n)(I - A_n)^T \tilde{y} = \tilde{b}, \quad \tilde{x} = (I - A_n)^T \tilde{y}, \quad (4.10)$$

with $(I - A_n)_n^T (I - A_n)$ and $(I - A_n)(I - A_n)^T$ are symmetric.

Now, in order to solve the system (4.8), we solve its two equivalent systems (4.9) and (4.10). We know that the matrices $(I - A_n)_n^T (I - A_n)$ and $(I - A_n)(I - A_n)^T$ do not have a small condition number. Thus, we apply conjugate gradient normal equation residual CGNR method to (4.9) and the conjugate gradient normal

equation error CGNE method to (4.10) with diagonal preconditioner D_n in order to obtain a new matrices with a smaller condition number.

More precisely, by applying the diagonal preconditioner to (4.9), we have then the following preconditioned system

$$D_n^{-1}(I - A_n)^T(I - A_n)\tilde{x} = D_n^{-1}(I - A_n)^T\tilde{b}, \quad (4.11)$$

with the condition number

$$k(D_n^{-1}(I - A_n)^T(I - A_n)) = k(D_n^{-1/2}(I - A_n)^T(I - A_n)D_n^{-1/2}) = O(1).$$

We apply again the diagonal preconditioner to (4.10), we get the following preconditioned system

$$D_n^{-1}(I - A_n)^T(I - A_n)\tilde{y} = D_n^{-1}\tilde{b}, \quad \tilde{x} = (I - A_n)^T\tilde{y}, \quad (4.12)$$

with the condition number

$$k(D_n^{-1}(I - A_n)(I - A_n)^T) = k(D_n^{-1/2}(I - A_n)(I - A_n)^TD_n^{-1/2}) = O(1).$$

Hence, we can solve the system (4.11) by applying the conjugate gradient normal equation residual CGNR method and (4.12) by applying the conjugate gradient normal equation error CGNE method which give as a linear convergence rate (see [9]).

Thus, the equivalent form of (4.11) is

$$\tilde{A}_n y_1 = z_1, \quad (4.13)$$

where

$$y_1 := D_n \tilde{x}, \quad z_1 := D_n^{-1}(I - A_n)^T \tilde{b},$$

and

$$\tilde{A}_n := D_n^{-1}(I - A_n)^T(I - A_n)D_n^{-1}.$$

The equivalent form of (4.12) is

$$\tilde{A}'_n y_2 = z_2, \quad (4.14)$$

where

$$y_2 = D_n \tilde{y}, \quad z_2 = D_n^{-1} \tilde{b},$$

and

$$\tilde{A}'_n = D_n^{-1}(I - A_n)(I - A_n)^TD_n^{-1}.$$

In each iteration of CGNR and CGNE methods, requires computing $(I - A_n)^T v_1$ and $(I - A_n) v_2$ for some vectors v_1 and v_2 respectively, and then solving (4.13) and (4.14) (see [25]).

Well, after some updates to CG method, we can solve the systems $D_n \tilde{x} = y_1$ and $D_n \tilde{y} = y_2$ respectively.

For solving the above systems, we use the algorithm presented in [9].

- For the case $(I - A_n)^T v_1$, since

$$(I - A_n) = W_n(I - T_n)W_n^{-1},$$

we get

$$(I - A_n)^T v_1 = (W_n^{-1})^T (I - T_n^T) u_1,$$

where $u_1 = W_n^T v_1$, and by using FWT we could then compute u_1 in $O(n)$ operations ([4,25]).

In addition, by using FFT we could then compute $(I - T_n)u_1$ in $O(n \log n)$ operations ([5,27]).

In the end, to solve $(I - A_n)v_1 = (W_n^{-1})^T (I - T_n)u_1$ we use FWT and Strang's algorithm given in [27]. Therefore, the operation cost decreased to $O(n \log n)$. Regarding the system $D_n \tilde{x} = y_1$ we just need $O(n)$ operations.

Hence, the cost per iteration for (4.9) is $O(n \log n)$.

- For the case $(I - A_n)v_2$, by the similar way as above, we get the cost per iteration for (4.10) is $O(n \log n)$.

Consequently, the total cost per iteration is $O(2n \log n)$.

Finally, we can solve the systems (4.7), (4.8) in $O(2n \log n)$, as a result of the independence of the iterations and n .

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References

1. A. Arikoglu, I. Ozkol, *Solutions of integral and integro-differential equation systems by using differential transform method*, Comput. Math. Appl. 56 (2008), 2411-2417.
2. B. Asady, M. Tavassoli Kajani, A. Hadi Vencheh, A. Heydari, *Solving second kind integral equations with hybrid Fourier and block-pulse functions*, Appl. Math. Comput. 160 (2005), 517-522.
3. Z. Avazzadeh, M. Heydari, G.B. Loghmani; *Numerical solution of Fredholm integral equations of the second kind by using integral mean value theorem*. App. Math. Model. 35 (2011), 2374-2383.
4. G. Beylkin, R. Coifman and V. Rokhlin, *Fast wavelet transforms and numerical algorithms I*, Comm. Pure Appl. Math. 44 (1991), 141-183.
5. R. Chan and G. Strang, *Toeplitz equations by conjugate gradients with circulant preconditioner*, SIAM J. Sci. Statist. Comput. 10 (1989), 104-119.
6. C. Chui, *An Introduction to Wavelet*, Academic Press, 1992.
7. I. Daubechies, *Orthonormal bases of compactly supported wavelets*, Comm. Pure Appl. Math. 41 (1988), 909-996.
8. A. Ghorbani, J. Saberi-Nadjafi, *Exact solutions for nonlinear integral equations by a modified homotopy perturbation method*, Comp. Math. appl. 56 (2008) 1032-1039.

9. G. Golub and C. Van Loan, *Matrix Computation*, 4th Edition, Johns Hopkins University Press, 2013.
10. B. Jawerth and W. Sweldens, *An Overview of Wavelet Based Multiresolution Analyses*, SIAM Rev., 36 (1994), 377-412.
11. X. Q. Jin and J. Y. Yuan, *A wavelet method for the first kind integral equations with kernel $k(x - y)$* , Taiwanese Journal of Mathematics, 2 (1998), 427-434.
12. A. Karoui, *Wavelets: Properties and approximate solution of a second kind integral equation*, Comput. Math. Appl. 46 (2003), 263-277.
13. D. T. L. Lee and A. Yamamoto, *Wavelet Analysis: Theory and Applications*, Hewlett-Packard Journal, 1994.
14. X.Z. Liang, M. C. Liu, X.J. Che, *Solving second kind integral equations by Galerkin methods with continuous orthogonal wavelets*. J. Comput. Appl. Math. 136 (2001), 149-161.
15. K. Maleknejad, N. Aghazadeh, R. Mollapourasl, *Numerical solution of Fredholm integral equation of the first kind with collocation method and estimation of error bound*. Appl. Math. Comput. 179 (2006), 352-359.
16. K. Maleknejad, T. Lotfi, K. Mahdiani, *Numerical solution of first kind Fredholm integral equations with wavelets-Galerkin method (WGM) and wavelets precondition*, Appl. Math. Comput. 186 (2007), 794-800.
17. K. Maleknejad, T. Lotfi, Y. Rostami, *Numerical computational method in solving Fredholm integral equations of the second kind by using Coifman wavelet*, Appl. Math. Comput. 186 (2007), 212-218.
18. K. Maleknejad, M. Tavassoli Kajani, *Solving second kind integral equations by Galerkin methods with hybrid Legendre and Block-Pulse functions*. Appl. Math. Comput. 145 (2003), 623-629.
19. K. Maleknejad, M. Yousefi, *Numerical solution of the integral equation of the second kind by using wavelet bases of Hermite cubic splines*, Appl. Math. Comput. 183 (2006), 134-141.
20. K. Maleknejad, M. Yousefi, K. Nouri, *Computational methods for integrals involving functions and Daubechies wavelets*, Appl. Math. Comput. 189 (2007), 1828-1840.
21. A. Mennouni, *A projection method for solving Cauchy singular integro-differential equations*, Applied Mathematics Letters 25 (2012), 986-989.
22. Y. Meyer, *Wavelets and Operators*, Cambridge University Press, 1992.
23. R. Piessens, *Computing integral transforms and solving integral equations using Chebyshev polynomial approximations*, J. Comput. Appl. Math. 121 (2000), 113-124.
24. J. Rashidinia, M. Zarebnia, *Numerical solution of linear integral equations by using sinc-collocation method*, Appl. Math. Comput. 168 (2005), 806-822.
25. Y. Saad, *Iterative Methods for Sparse Linear Systems*, Society for Industrial and Applied Mathematics, 2003.
26. C. Sidney Burrus, Ramesh A. Gopinath, and Haitao Guo, *Introduction to Wavelets and Wavelet Transforms*, A Primer, Prentice Hall Upper Saddle River, New Jersey 1998.
27. G. Strang, *A proposal for Toeplitz matrix calculations*, Stud. Appl. Math. 74 (1986), 171-176.
28. P. Zhang and Y. Zhang, *Wavelet method for boundary integral equation*, Journal of Computational Mathematics, 18 (2000), 25-42.

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