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Fast Wavelet Collocation Methods for Second Kind Integral Equations on Polygons

Yi Wang

Dissertation submitted to the College of Arts and Sciences at West Virginia University in partial fulfillment of the requirements for the degree of

> Doctor of Philosophy in Mathematics

Yuesheng Xu, Ph.D., Chair Martina E. Bachlechner, Ph.D. Harvey Diamond, Ph.D. Weifu Fang, Ph.D. Sherman D. Riemenschneider, Ph.D.

> 2003 Morgantown, West Virginia

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Fast Wavelet Collocation Methods for Second Kind Integral Equations on Polygons

Yi Wang

(ABSTRACT)

In this thesis we develop fast wavelet collocation methods for integral equations of the second kind with weakly singular kernels over *polygons*. For this purpose, we construct *multiscale* wavelet functions and collocation functionals having vanishing moments. Moreover, we propose several truncation strategies, which lead to fast algorithms, for the coefficient matrix of the corresponding discrete system. Critical issues for numerical implementation of such methods are considered, such as choices of practical truncation strategies, numerical integration of weakly singular integrals, error controls of numerical quadrature and numerical solutions of resulting compressed linear systems. Numerical experiments are given to demonstrate proposed ideas and methods. Finally, parallel computing using developed methods is investigated.

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Chapter 1

Introduction

In this thesis we develop fast wavelet collocation methods for solving Fredholm integral equations of the second kind on *polygons*. We consider integral equations of the form

$$u(t) - \int_{E} K(s,t) \ u(s) \ ds = f(t), \quad t \in E,$$
 (1.1)

where E is a polygon in \mathbb{R}^2 , and the kernel K(s,t) is a weakly singular function on $E \times E$. Introducing an operator

$$(\mathcal{K}u)(t) = \int_E K(s,t)u(s)ds, \ t \in E$$

(1.1) can be written into

$$(\mathcal{I} - \mathcal{K})u = f. \tag{1.2}$$

This type of equations has many important applications including boundary integral equations [At1, At2] and radiosity equations [ACr, AC2]. There have been many theoretical studies as well as numerical methods on such Fredholm integral equations of the second kind (cf. [At2, K]).

Using a conventional method such as Galerkin method or collocation method to discretize an integral equation will lead to a linear system with a full coefficient matrix. It's computationally costly to generate the full coefficient matrix. Especially, when higher dimensional integral equations are considered, the computational cost to generate the full coefficient matrix is formidable. Therefore fast algorithms are extremely crucial for numerical solutions of integral equations. Recently, wavelet methods have been playing important roles in seeking fast algorithms for numerical solutions of integral equations.

Wavelet Galerkin methods and Petrov-Galerkin methods were investigated recently by [Al, BCR, CMX1, CMX2, DPS1, DPS2, MXZ, PS, PSS, R] as well as references cited therein. A fast multilevel method was developed in [CMX4] using a multilevel decomposition of the approximation space. [DPS2, MXZ] studied compression strategies with slightly different focuses. [DPS2] studied wavelet Galerkin methods using periodic wavelets based on refinement equations for periodic problems, while [MXZ] developed wavelet Galerkin methods using piecewise polynomial wavelets. One advantage of using piecewise polynomial wavelets is that the wavelet functions have close forms which provide convenience for computation. The implementation of wavelet Galerkin methods based on refinement equations was done in [DKPS]. The implementation of wavelet Galerkin methods use L_2 analysis and therefore the vanishing moments of the multiscale basis functions naturally lead to matrix truncation schemes.

Collocation methods due to lower computational cost for evaluations of integrals (for example , see [At1, AC1, CX]), receive more favorable attention from engineering fields. Yet less attention has been paid to them. For collocation methods, the appropriate space to work in is L_{∞} and, this causes challenging technical obstacles for identifying good matrix truncation strategies. The recent major work in this area is done in [CMX3]. [CMX3] provides a theoretical framework of fast wavelet collocation methods for solving integral equations on invariant domains associated with families of contractive mappings. [CWX2] studies the actual computation of 1-D integral equations using methods proposed in [CMX3].

Based on the framework discussion in [CMX3], it is the purpose of this thesis to investigate actual computing methods for 2-D wavelet collocation methods. By constructing semi-biorthogonal wavelets and collocation functionals for 2-D collocation scheme, practical matrix truncation strategies are studied. Theses practical strategies lead to fast algorithms for solving integral equations. The 2-D wavelets are constructed on invariant domains and have simple expressions. Both wavelet basis functions and associated collocation functionals have vanishing moments of certain degrees and multilevel structures. The collocation functionals are constructed through set wavelets. Results from the truncated system have the same accuracy and convergence order as those from uncompressed ones, yet the computing cost is dramatically reduced from $O(N^2)$ to $O(N \log N)$, where N is the dimension of the approximation space. The approximation solution has a convergence order of $O(N^{k/2} \log N)$, where k is the total degree of piecewise polynomials in the approximation space. The condition number of the compressed coefficient matrix is of order $O(\log^2 N)$. The methods developed in this thesis have application not only to the integral equations, but also can be used to solve certain 3-D boundary value problems.

More specifically, In chapter 2 we briefly review the methods proposed in [CMX3]. In Chapter 3, we develop our wavelet collocation methods on the unit triangle. The multilevel wavelet basis and collocation functionals on the unit triangle are constructed. Properties of the constructed wavelets and collocation functionals are also investigated. A reconstruction algorithm of the sought function is also discussed in this chapter. In Chapter 3, we discuss how to compute the coefficient matrix in detail. Other critical issues are considered too, such as, compressing schemes, computation of integrals with singular kernels, and solutions of the resulting linear system. We provide two *practical* truncation strategies based on the theoretical truncation strategy proposed in [CMX3]. An efficient adaptive quadrature rule for computing singular double integrals is also proposed. The control of errors introduced from the numerical integration is analyzed. Moreover, we develop a multilevel iterative method solving resulting sparse linear systems by adapting ideas proposed in [FMX]. In Chapter 5, fast wavelet methods on polygons are developed based on methods on the unit triangle and triangulations of polygons. In Chapter 6, we combine all proposed methods together to solve a number of examples to demonstrate accuracies and efficiencies of proposed methods. Parallel computing algorithms and their implementation in a Beowulf cluster are also investigated and demonstrated.

Chapter 2

Fast Wavelet Collocation Methods

2.1 Multiscale Collocation Schemes

In this chapter we briefly review the methods developed in [CMX3] for solving Fredholm integral equations of the second kind. Let E be a compact set with star shape in the ddimensional Euclidean space \mathbb{R}^d , and let $\mathbb{X} = L^{\infty}(E)$, $\mathbb{V} = C(E)$. Suppose that K is a weakly singular kernel, that is , for every $s \in E$, $K(s, .) \in L^1(E)$. Therefore, the operator $\mathcal{K} : \mathbb{X} \to \mathbb{V}$ defined by

$$(\mathcal{K}u)(s) := \int_E K(s,t)u(t)dt, \ s \in E$$
(2.1)

is compact in X. We consider Fredholm integral equations of the second kind

$$u - \mathcal{K}u = f, \tag{2.2}$$

where $f \in \mathbb{X}$ is a given function and $u \in \mathbb{X}$ is the unknown to be determined. By Fredholm alternative theorem, when the corresponding homogeneous equation has only trivial solution, equation (2.2) has a unique solution in \mathbb{X} .

Let the index set $\mathbb{N}_0 := \{0, 1, \ldots\}$. Assume that a sequence of approximation spaces $\mathbb{F}_n, n \in \mathbb{N}_0$ of X, satisfying

$$\mathbb{F}_n \subseteq \mathbb{F}_{n+1}, \quad n \in \mathbb{N}_0 \tag{2.3}$$

and

$$\mathbb{V} \subseteq \overline{\bigcup_{n \in \mathbb{N}_0} \mathbb{F}_n}.$$
(2.4)

By the nestedness (2.3) \mathbb{F}_n can be further decomposed into the direct sum of a family of spaces

$$\mathbb{F}_n = \mathbb{W}_0 \oplus \mathbb{W}_1 \oplus \ldots \oplus \mathbb{W}_n. \tag{2.5}$$

where spaces \mathbb{W}_n have multiscale structures and will be constructed as piecewise polynomial functions on E. Note $\mathbb{F}_0 = \mathbb{W}_0$. Let $w(n) := \dim \mathbb{W}_n$ and, we have that

$$f(n) := \dim \mathbb{F}_n = \sum_{r \in \mathbb{Z}_{n+1}} w(r), \qquad (2.6)$$

where, $\mathbb{Z}_m := \{0, 1, \dots, m-1\}$. A multiscale partition of the set E is also needed for the wavelet collocation method. It consists of a family of partitions $\{E_n : n \in \mathbb{N}_0\}$ of E. For each $n \in \mathbb{N}_0$ the partition E_n consists of a family of subsets $\{E_{ni} : i \in \mathbb{Z}_{e(n)}, e(n) = cw(n), n \ge 0\}$ of E with the properties

$$\operatorname{meas}(E_{ni} \cap E_{ni'}) = 0, \quad i, i' \in \mathbb{Z}_{e(n)}, \ i \neq i',$$

$$(2.7)$$

and

$$\bigcup_{i \in \mathbb{Z}_{e(n)}} E_{ni} = E.$$
(2.8)

Where c is a generic constant in this thesis. (We sometimes also use C for the generic constant). When n = 0 the set E is the only cell in the partition. See Fig. 2.1 for an example of partitions when E is the unit triangle. Further, the partition E_n has the following property. Namely, for some positive constants c_- , c_+ and all $n \in \mathbb{N}_0$

$$c_{-}\mu^{-n/d} \le \max\{d(E_{n,i}) : i \in \mathbb{Z}_{e(n)}\} \le c_{+}\mu^{-n/d}.$$
 (2.9)

where d is the spatial dimension, and μ is a constant positive integer. For a set $A \subset \mathbb{R}^d$, d(A) represents the diameter of A, i.e.,

$$d(A) := \sup\{|x - y| : x, y \in A\},$$
(2.10)

and |.| denotes the Euclidean norm on the space \mathbb{R}^d . The family of partitions $\{E_n : n \in \mathbb{N}_0\}$ is used to specify supports of both basis functions and linear functionals. Specifically, assume

$$\mathbb{W}_n := \operatorname{span} W_n, \quad n \in \mathbb{N}_0, \tag{2.11}$$



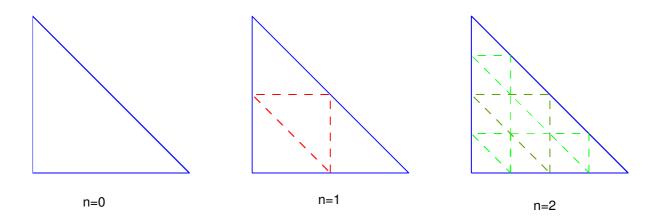


Figure 2.1: Multiscale partitions of the unit triangle Ω

where, $W_n := \{w_{nm} : m \in \mathbb{Z}_{w(n)}\}$. The basis function w_{nm} are required to be locally supported and their supports are shrinking as level *n* increases. Denote the support of the function w_{nm} by S_{nm} . S_{nm} is contained in E_{ni} for some $i \in \mathbb{Z}_{e(n)}$. On the other hand, a set of linear functionals in \mathbb{V}^* given by

$$L_n := \{ \ell_{nm} : m \in \mathbb{Z}_{w(n)} \}, \quad n \in \mathbb{N}_0.$$
(2.12)

is introduced. Where, \mathbb{V}^* is the dual space of \mathbb{V} , and the linear functional ℓ_{nm} is a finite sum of point evaluations

$$\ell_{nm} = \sum_{s \in \hat{E}_{nm}} c_s \delta_s, \tag{2.13}$$

where c_s are constants and \hat{E}_{nm} is considered as the "support" of the functional ℓ_{nm} . \hat{E}_{nm} is a finite subset of distinct points in E_{nm} with its cardinality bounded independent of $n \in \mathbb{N}$. For any $s \in E$, δ_s denotes the linear functional in \mathbb{V}^* defined for $v \in \mathbb{V}$ by the equation $\langle \delta_s, v \rangle = v(s)$. We shall need to evaluate δ_s on functions in X. Therefore, as in [AGS] we take any norm preserving extension of δ_s to X and use the same notation for the extension. In particular, this extension allows us to evaluate piecewise polynomials on E.

The linear functionals and multiscale basis functions are constructed such that, for any $n, n' \in \mathbb{N}_0$

$$\langle \ell_{n'm'}, w_{nm} \rangle = \delta_{nn'} \delta_{mm'}, \quad (n, m), (n', m') \in \mathbb{U}, \quad n \le n', \tag{2.14}$$

where $\mathbb{U} := \{(i, j) : i \in \mathbb{N}_0, j \in \mathbb{Z}_{w(i)}\}$ and $\delta_{ii'}$ is the Kronecker delta. Moreover, for some positive constant γ ,

$$\sum_{m \in \mathbb{Z}_{w(n)}} |\langle \ell_{n'm'}, w_{nm} \rangle| \le \gamma, \quad (n,m), (n',m') \in \mathbb{U}, \quad n > n'.$$
(2.15)

To solve (2.2) using collocation scheme, a vector $\mathbf{u}_n := [u_{ij} : (i, j) \in \mathbb{U}_n]$, where \mathbb{U}_n is the set of lattice points in \mathbb{R}^2 defined as $\{(i, j) : j \in \mathbb{Z}_{w(i)}, i \in \mathbb{Z}_{n+1}\}$, is sought such that the function

$$u_n := \sum_{(i,j)\in\mathbb{U}_n} u_{ij} w_{ij} \tag{2.16}$$

in \mathbb{F}_n satisfies

$$\langle \ell_{i'j'}, u_n - \mathcal{K}u_n \rangle = \langle l_{i'j'}, f \rangle, \ (i', j') \in \mathbb{U}_n.$$
 (2.17)

Equivalently, a linear system of equations

$$(\mathbf{E}_n - \mathbf{K}_n)\mathbf{u}_n = \mathbf{f}_n,\tag{2.18}$$

is obtained, where

$$\mathbf{K}_n := [\langle \ell_{i'j'}, \mathcal{K}w_{ij} \rangle]_{f(n) \times f(n)}, \qquad (2.19)$$

$$\mathbf{E}_n := [\langle \ell_{i'j'}, w_{ij} \rangle]_{f(n) \times f(n)}, \qquad (2.20)$$

and

$$\mathbf{f}_n := [\langle l_{i'j'}, f \rangle : (i', j') \in \mathbb{U}_n].$$
(2.21)

2.2 General Construction of Multiscale Functions and Functionals

We start with a positive integer μ and a family of contractive affine mappings on $\mathbb{R}^d \Phi := \{\phi_e : e \in Z_\mu\}$. It is known (see, [H]) that there exists a unique compact subset $E \in \mathbb{R}^d$ such that

$$\Phi(E) = E, \tag{2.22}$$

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where

$$\Phi(E) := \bigcup_{e \in Z_{\mu}} \phi_e(E).$$
(2.23)

The set E is called the invariant set associated with the family of mappings Φ . Additional restriction to the family of mappings Φ is made:

- (a) For every $e \in \mathbb{Z}_{\mu}$, ϕ_e^{-1} is continuous on E;
- (b) The set E has non-empty interior and

$$\operatorname{meas}(\phi_e(E) \cap \phi_{e'}(E)) = 0, \quad e, e' \in \mathbb{Z}_{\mu}, \ e \neq e'.$$

$$(2.24)$$

 Φ is used to obtain multiscale partitions $\{E_n : n \in \mathbb{Z}_0\}$ of the set E. To this end, let $\mathbf{e} := (e_0, e_1, \dots, e_{n-1}) \in \mathbb{Z}_{\mu}^n := \mathbb{Z}_{\mu} \times \dots \times \mathbb{Z}_{\mu}$. Define a composite mapping

$$\phi_{\mathbf{e}} := \phi_{e_0} \circ \phi_{e_1} \circ \ldots \circ \phi_{e_{n-1}} \tag{2.25}$$

and a number

$$\mu(\mathbf{e}) := \mu^{n-1} e_0 + \ldots + \mu e_{n-2} + e_{n-1}.$$
(2.26)

Since E is an invariant set with respect to Φ and conditions (a) with (b) it follows that the collection of sets

$$E_n := \{ E_{n,\mathbf{e}} : E_{n,\mathbf{e}} = \phi_{\mathbf{e}}(E), \ \mathbf{e} \in \mathbb{Z}^n_\mu \}$$

$$(2.27)$$

form a partition of E. Let \mathbb{F}_n be the spaces of all functions such that their restriction to any cell $E_{n,\mathbf{e}}$, $\mathbf{e} \in \mathbb{Z}^n_{\mu}$ is a polynomial of total degree $\leq k - 1$, $k \geq 1$. Therefore

$$m := \dim \mathbb{F}_0 = \begin{pmatrix} k+d-1\\ d \end{pmatrix}.$$
(2.28)

Let

$$G_0 = \{ t_j : t_j \in E, \text{ and } t_j \neq t_i, \text{ for } i \neq j, i, j \in \mathbb{Z}_m \},$$

$$(2.29)$$

which is refinable relative to the mappings Φ , that is

$$G_0 \subset \Phi(G_0). \tag{2.30}$$

Set

$$G_1 := \Phi(G_0), \quad V_1 := G_1 \setminus G_0 = \{t_{m+j} : j \in \mathbb{Z}_r\}$$
 (2.31)

with $r := (\mu - 1)m$. Now, Assume

$$\mathbb{F}_0 := \operatorname{span}\{\psi_j : j \in \mathbb{Z}_m\},\tag{2.32}$$

and $\psi_0, \psi_1, \ldots, \psi_{m-1}$ satisfy unique Lagrange interpolation conditions

$$\psi_i(t_j) = \delta_{ij}, \quad i, j \in \mathbb{Z}_m. \tag{2.33}$$

A construction of refinable points $\{t_j : j \in \mathbb{Z}_m\} \in E$ that admits a unique *d*-dimensional Lagrange interpolation is presented in [MSX1]. Introduce linear operators $\mathcal{T}_e : \mathbb{X} \to \mathbb{X}, e \in \mathbb{Z}_{\mu}$ defined by

$$(T_e x)(t) := x(\phi_e^{-1}(t))\chi_{\phi_e(E)}(t), \qquad (2.34)$$

where χ_S denote the characteristic function of the set S. Therefore it follows that

$$\mathbb{F}_n = \bigoplus_{e \in Z_\mu} \mathcal{T}_e \mathbb{F}_{n-1}, \quad n \in \mathbb{Z},$$
(2.35)

where $\mathbb{A} \oplus \mathbb{B}$ denotes the direct sum of the spaces \mathbb{A} and \mathbb{B} . The functions $\psi_{m+j} \in \mathbb{F}_1$ such that $\psi_{m+j}(t_{m+i}) = \delta_{ij}$, $t_{m+i} \in V_1$, $i, j \in \mathbb{Z}_r$ with $\psi_j \in \mathbb{F}_0$, $j \in \mathbb{Z}_m$, defined in (2.33) form a basis for \mathbb{F}_1 . This basis doesn't have vanishing moment. So instead another basis for \mathbb{F}_1 consisting of functions with vanishing moments is constructed. For this purpose, set

$$w_{0j} := \psi_j, \quad j \in \mathbb{Z}_m \tag{2.36}$$

let q := m + r and for $j \in \mathbb{Z}_r$ find a vector $[c_{js} : s \in \mathbb{Z}_q] \in \mathbb{R}^q$ such that

$$w_{1j} := \sum_{s \in \mathbb{Z}_q} c_{js} \psi_s, \quad j \in \mathbb{Z}_r$$
(2.37)

satisfies

$$(w_{1j}, w_{0j'}) = 0, \quad j' \in \mathbb{Z}_m, \ j \in \mathbb{Z}_r,$$
 (2.38)

where (,) is the usual L^2 inner product of two functions on E. (2.38) leads to a linear system of rank m with m equations and q unknowns, thus it has r linearly independent solutions. Denote them by $w_{1j}, j \in \mathbb{Z}_r$. These functions form a basis for the space \mathbb{W}_1 . To construct a basis for $\mathbb{W}_i, i \geq 2$, for $\mathbf{e} := (e_0, \ldots, e_{n-1}) \in \mathbb{Z}_{\mu}^n$ introduce a composite operator $\mathcal{T}_{\mathbf{e}}$ by

$$\mathcal{T}_{\mathbf{e}} := \mathcal{T}_{e_0} \circ \ldots \circ \mathcal{T}_{e_{n-1}}.$$
(2.39)

For i = 2, 3, ..., n, let

$$w_{ij} := \mathcal{T}_{\mathbf{e}} w_{1l}, \quad j = \mu(\mathbf{e})r + l, \ \mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}, \ l \in \mathbb{Z}_r$$

$$(2.40)$$

and

$$\mathbb{W}_i := \operatorname{span}\{w_{ij} : j \in \mathbb{Z}_{w(i)}\}.$$
(2.41)

Note that the support of w_{ij} is contained in $S_{ij} := \phi_{\mathbf{e}}(E), j \in \mathbb{Z}_{w(i)}$.

The collocation functionals are defined in a similar manner. Define

$$\ell_{0j} := \delta_{t_j}, \quad j \in \mathbb{Z}_m \tag{2.42}$$

and for $j' \in \mathbb{Z}_r$, find the vector $[c'_{j's} : s \in \mathbb{Z}_q]$ such that

$$\ell_{1j'} := \sum_{s \in \mathbb{Z}_q} c'_{j's} \delta_{t_s}, \quad j' \in \mathbb{Z}_r$$
(2.43)

satisfies equations

$$\langle \ell_{1j'}, w_{0j} \rangle = 0, \quad j \in \mathbb{Z}_m, \quad j' \in \mathbb{Z}_r$$

$$(2.44)$$

and

$$\langle \ell_{1j'}, w_{1j} \rangle = \delta_{jj'}, \quad j \in \mathbb{Z}_r, \quad j' \in \mathbb{Z}_r.$$
 (2.45)

(2.44) and (2.45) lead to a $q \times q$ linear system of equations. The coefficient matrix for this linear system of equations is

$$\mathbf{A} := \left[\left\langle \delta_{t_{i'j'}}, w_{ij} \right\rangle \right]_{(i,j), (i'j') \in \mathbb{U}_1}.$$
(2.46)

A is proved to be nonsingular in [CMX3]. To generate a multiscale collocation functionals, introduce for any $e \in \mathbb{Z}_{\mu}$ a linear operator $\mathcal{L}_e: V^* \to V^*$ defined by the equation

$$\langle \mathcal{L}_e \ell, v \rangle = \langle \ell, v \circ \phi_e \rangle, \quad v \in \mathbb{X}, \ \ell \in \mathbb{X}^*,$$

$$(2.47)$$

Moreover, for $\mathbf{e} := (e_0, \ldots, e_{n-1}) \in \mathbb{Z}_{\mu}^n$, define a composite operator

$$\mathcal{L}_{\mathbf{e}} := \mathcal{L}_{e_0} \circ \ldots \circ \mathcal{L}_{e_{n-1}}.$$
 (2.48)

Consequently, for any $\mathbf{e}, \mathbf{e}' \in \mathbb{Z}^i_{\mu}$, $w \in \mathbb{X}$ and $\ell \in \mathbb{X}^*$, we have that

$$\langle \mathcal{L}_{\mathbf{e}}l, \mathcal{T}_{\mathbf{e}'}w \rangle = \langle l, w \rangle \delta_{\mathbf{e}\mathbf{e}'}.$$
 (2.49)

In addition, for i > 1, $j = \mu(\mathbf{e})r + l$, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $l \in \mathbb{Z}_r$ we define

$$\ell_{ij} := \mathcal{L}_{\mathbf{e}} \ell_{1l}. \tag{2.50}$$

Note that

$$\langle \ell_{ij}, v \rangle = \langle \ell_{1l}, v \circ \phi_{\mathbf{e}} \rangle = \sum_{s \in \mathbb{Z}_q} c'_{ls} v(\phi_{\mathbf{e}}(t_s)).$$
 (2.51)

Observe that the "support" of ℓ_{ij} is also contained in S_{ij} .

Chapter 3

Wavelets and Collocation Functionals on the Unit Triangle

Our following discussion is on the unit triangle $\Omega := \{(x, y) | 0 \le x \le 1, 0 \le y \le 1 - x\}$. We first introduce a family of contractive mappings $\Phi = \{\phi_e, e \in \mathbb{Z}_4\}$ to subdivide the domain Ω . Define

$$\phi_{0}(x,y) = \left(\frac{x}{2}, \frac{y}{2}\right),
\phi_{1}(x,y) = \left(\frac{x+1}{2}, \frac{y}{2}\right),
\phi_{2}(x,y) = \left(\frac{x}{2}, \frac{y+1}{2}\right).
\phi_{3}(x,y) = \left(\frac{1-x}{2}, \frac{1-y}{2}\right).$$
(3.1)

These four mappings subdivide the unit triangle Ω into four small triangles Ω_0 , Ω_1 , Ω_2 , and Ω_3 (see Fig. 3.1), with $\Omega_i = \phi_i(\Omega)$, $i \in \mathbb{Z}_4$. We then can define four operators \mathcal{T}_i , $i \in \mathbb{Z}_4$ through the four mappings ϕ_i , $i \in \mathbb{Z}_4$, namely,

$$(\mathcal{T}_i \circ g)(\xi, \eta) = g(\phi_i^{-1}(\xi, \eta))\chi_{\Omega_i}(\xi, \eta), \quad i \in \mathbb{Z}_4,$$
(3.2)

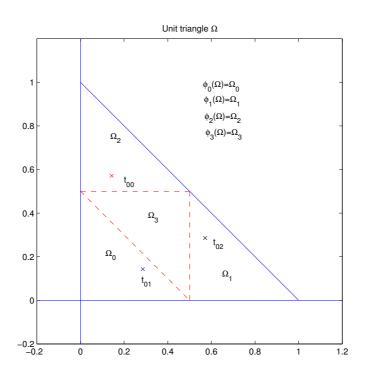


Figure 3.1: Original domain–unit triangle Ω

where,

$$\phi_0^{-1}(x,y) = (2x,2y),$$

$$\phi_1^{-1}(x,y) = (2x-1,2y),$$

$$\phi_2^{-1}(x,y) = (2x,2y-1),$$

$$\phi_3^{-1}(x,y) = (1-2x,1-2y).$$
(3.3)

We choose as the original collocation points $t_{00} = (\frac{1}{7}, \frac{4}{7})$, $t_{01} = (\frac{2}{7}, \frac{1}{7})$, and $t_{02} = (\frac{4}{7}, \frac{2}{7})$ (see Fig. 3.1). Note the set $G_0 := \{t_{0j} : j \in \mathbb{Z}_3\}$ is refinable with respect to the family of mappings $\Phi := \{\phi_e : e \in \mathbb{Z}_4\}$.

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3.1 The Approximation Spaces

Let \mathbb{W}_0 be a subspace of \mathbb{X} with dim $\mathbb{W}_0 = 3$. Assume \mathbb{W}_0 is a space of linear polynomials on Ω . Moreover assume

$$\mathbb{W}_0 = \operatorname{span}\{w_{0j}, \ j \in \mathbb{Z}_3\},\tag{3.4}$$

where, w_{0j} , $j \in \mathbb{Z}_3$ are determined by interpolating conditions

$$w_{0j}(t_{0k}) = \delta_{jk}, \quad j,k \in \mathbb{Z}_3. \tag{3.5}$$

To find w_{0j} , $j \in \mathbb{Z}_3$, write $w_{0j} = b_{j_0}x + b_{j_1}y + b_{j_2}$, $j \in \mathbb{Z}_3$. Define a matrix $\mathbf{B} = [b_{jk}] : j, k \in \mathbb{Z}_3$, $\mathbf{B} \in M_{3,3}$ (throughout this thesis $M_{m,n}$ denotes a $m \times n$ matrix over the field \mathbb{R}). Let

$$\mathbf{A} = \begin{bmatrix} \frac{1}{7} & \frac{4}{7} & 1\\ \frac{2}{7} & \frac{1}{7} & 1\\ \frac{4}{7} & \frac{2}{7} & 1 \end{bmatrix}.$$
 (3.6)

Solve

$$\mathbf{AB} = \mathbf{I} \tag{3.7}$$

for **B**, where, **I** is the identity matrix in $M_{3,3}$. We obtain

$$\mathbf{B} = \begin{bmatrix} -1 & -2 & 3\\ 2 & -3 & 1\\ 0 & 2 & -1 \end{bmatrix}.$$
 (3.8)

Thus,

$$w_{00}(x, y) = -x + 2y,$$

$$w_{01}(x, y) = -2x - 3y + 2,$$

$$w_{02}(x, y) = 3x + y - 1.$$
(3.9)

With $\mathbb{W}_0(=\mathbb{F}_0)$, we can find

$$\mathbb{F}_1 = \bigoplus_{j \in \mathbb{Z}_4} \mathcal{T}_j \mathbb{W}_0. \tag{3.10}$$

Let \mathbb{W}_1 be the orthogonal complement of \mathbb{W}_0 in \mathbb{F}_1 , and define

$$\mathbb{W}_n = \bigoplus_{j \in \mathbb{Z}_4} \mathcal{T}_j \mathbb{W}_{n-1}, \ n \ge 2, \ n \in \mathbb{Z}.$$
(3.11)

We can easily find that

$$\operatorname{Dim}\mathbb{F}_n = 3 \cdot 4^n, n \in \mathbb{N}_0, \tag{3.12}$$

and

$$\text{Dim}\mathbb{W}_n = 3^2 \cdot 4^{n-1}, \ n \ge 1, \ n \in \mathbb{Z}.$$
 (3.13)

3.2 Initial Wavelets in \mathbb{W}_1

We proceed in this section to construct a wavelet basis having vanishing moment of \mathbb{W}_1 . Now,

$$\begin{aligned} \mathcal{T}_{0} \circ w_{00} &= (-2x + 4y)\chi_{\Omega_{0}}, \\ \mathcal{T}_{0} \circ w_{01} &= (-4x - 6y + 2)\chi_{\Omega_{0}}, \\ \mathcal{T}_{0} \circ w_{02} &= (6x + 2y - 1)\chi_{\Omega_{0}}, \\ \mathcal{T}_{1} \circ w_{00} &= (-2x + 4y + 1)\chi_{\Omega_{1}}, \\ \mathcal{T}_{1} \circ w_{01} &= (-4x - 6y + 4)\chi_{\Omega_{1}}, \\ \mathcal{T}_{1} \circ w_{02} &= (6x + 2y - 4)\chi_{\Omega_{1}}, \\ \mathcal{T}_{2} \circ w_{00} &= (-2x + 4y - 2)\chi_{\Omega_{2}}, \\ \mathcal{T}_{2} \circ w_{01} &= (-4x - 6y + 5)\chi_{\Omega_{2}}, \\ \mathcal{T}_{2} \circ w_{02} &= (6x + 2y - 2)\chi_{\Omega_{2}}, \\ \mathcal{T}_{3} \circ w_{00} &= (2x - 4y + 1)\chi_{\Omega_{3}}, \\ \mathcal{T}_{3} \circ w_{01} &= (4x + 6y - 3)\chi_{\Omega_{3}}, \\ \mathcal{T}_{3} \circ w_{02} &= (-6x - 2y + 3)\chi_{\Omega_{3}}. \end{aligned}$$
(3.14)

Note $\mathcal{T}_0 w_{02}(t_{01}) = \mathcal{T}_1 w_{00}(t_{02}) = \mathcal{T}_2 w_{01}(t_{00}) = 1$. Write

$$\mathbf{S} = [w_{00}, w_{01}, w_{02}, \mathcal{T}_0 w_{00}, \mathcal{T}_0 w_{01}, \mathcal{T}_1 w_{01}, \mathcal{T}_1 w_{02}, \mathcal{T}_2 w_{00}, \mathcal{T}_2 w_{02}, \mathcal{T}_3 w_{00}, \mathcal{T}_3 w_{01}, \mathcal{T}_3 w_{02}]^T$$

$$:= [\psi_0, \psi_1, \psi_2, \psi_3, \psi_4, \psi_5, \psi_6, \psi_7, \psi_8, \psi_9, \psi_{10}, \psi_{11}]^T.$$
(3.15)

Define the wavelet basis functions of \mathbb{W}_1 by w_{1j} , $j \in \mathbb{Z}_9$. They are supported on Ω . Let $S := \{\psi_j, j \in \mathbb{Z}_{12}\}$. Since S is a set of basis functions of \mathbb{F}_1 , we have

$$w_{1j} = \sum_{s \in \mathbb{Z}_{12}} c_{js} \psi_s.$$
 (3.16)

To make $w_{1j}, j \in \mathbb{Z}_9$ have vanishing moment, require that

$$(w_{1k}, x^i y^j) = 0 \quad k \in \mathbb{Z}_9,$$
 (3.17)

where, $i, j \in \mathbb{Z}_2$, $i + j \leq 1$. Observe that

$$\int_{\Omega_0} f(x,y) dS = \int_0^{\frac{1}{2}} \int_0^{-x+\frac{1}{2}} f(x,y) dy dx,$$

$$\int_{\Omega_1} f(x,y) dS = \int_{\frac{1}{2}}^{1} \int_0^{-x+1} f(x,y) dy dx,$$

$$\int_{\Omega_2} f(x,y) dS = \int_0^{\frac{1}{2}} \int_{\frac{1}{2}}^{-x+1} f(x,y) dy dx,$$

$$\int_{\Omega_3} f(x,y) dS = \int_0^{\frac{1}{2}} \int_{-x+\frac{1}{2}}^{\frac{1}{2}} f(x,y) dy dx.$$
(3.18)

(3.17) yields the following linear system

$$\mathbf{Ac} = 0, \tag{3.19}$$

where,

$$\mathbf{A} = \begin{bmatrix} \frac{1}{6} & \frac{1}{6} & \frac{1}{24} \\ 0 & \frac{1}{24} & \frac{1}{8} & 0 & \frac{1}{192} & \frac{5}{192} & \frac{7}{192} & 0 & \frac{1}{64} & \frac{1}{48} & \frac{1}{64} & \frac{1}{192} \\ \frac{1}{8} & 0 & \frac{1}{24} & \frac{1}{64} & 0 & 0 & \frac{1}{192} & \frac{7}{192} & \frac{5}{192} & \frac{1}{192} & \frac{1}{48} & \frac{1}{64} \end{bmatrix}$$
(3.20)

and

$$\mathbf{c} = [c_0 \ c_1 \ c_2 \ c_3 \ c_4 \ c_5 \ c_6 \ c_7 \ c_8 \ c_9 \ c_{10} \ c_{11}]^T.$$
(3.21)

The null space of above linear system has dimension equal to 9, and is spanned by the set $\{\mathbf{w}_i : i \in \mathbb{Z}_9\}$, where \mathbf{w}_i corresponds to the *i*th row vector of the following 9×12 matrix

Now we can use matrix form to express the nine wavelet basis functions in the space \mathbb{W}_1 . Let **W** be the vector of the nine wavelets, i.e.,

$$\mathbf{W} = [w_{10}, w_{11}, w_{12}, w_{13}, w_{14}, w_{15}, w_{16}, w_{17}, w_{18}]^T.$$
(3.23)

For any nonsingular matrix $\mathbf{D} \in M_{9,9}$,

$$\mathbf{W} = \mathbf{D} \cdot \mathbf{C} \cdot \mathbf{S} \tag{3.24}$$

is a suitable basis for \mathbb{W}_1 satisfying (3.17). viz,

$$w_{1i} = \sum_{j \in Z_9} d_{ij} \sum_{k \in \mathbb{Z}_{12}} c_{jk} \psi_k.$$
(3.25)

We can take D = diag(1, 1, 1, 1, 1, 1, 1, 1, 1) and obtain

$$\begin{split} w_{10}(x,y) &= \frac{1}{8}(-2+3x+y)(\chi_{\Omega_0}+\chi_{\Omega_1}+\chi_{\Omega_2}) &+ \frac{1}{8}(22-45x-15y)\chi_{\Omega_3}, \\ w_{11}(x,y) &= \frac{1}{8}(1-2x-3y)(\chi_{\Omega_0}+\chi_{\Omega_1}+\chi_{\Omega_2}) &+ \frac{1}{8}(-23+30x+45y)\chi_{\Omega_3}, \\ w_{12}(x,y) &= \frac{1}{8}(-1-x+2y)(\chi_{\Omega_0}+\chi_{\Omega_1}+\chi_{\Omega_2}) &+ \frac{1}{8}(7+15x-30y)\chi_{\Omega_3}, \\ w_{13}(x,y) &= \frac{1}{8}(2-3x-5y)(\chi_{\Omega_0}+\chi_{\Omega_1}+\chi_{\Omega_3}) &+ \frac{1}{8}(-14+45x+11y)\chi_{\Omega_2}, \\ w_{14}(x,y) &= \frac{1}{8}(1+x-6y)(\chi_{\Omega_0}+\chi_{\Omega_1}+\chi_{\Omega_3}) &+ \frac{1}{8}(-15-15x+26y)\chi_{\Omega_2}, \\ w_{15}(x,y) &= \frac{1}{8}(2-7x-y)(\chi_{\Omega_0}+\chi_{\Omega_2}+\chi_{\Omega_3}) &+ \frac{1}{8}(-30+41x+15y)\chi_{\Omega_1}, \\ w_{16}(x,y) &= \frac{1}{8}(-1-2x+3y)(\chi_{\Omega_0}+\chi_{\Omega_2}+\chi_{\Omega_3}) &+ \frac{1}{8}(31-34x-45y)\chi_{\Omega_1}, \\ w_{17}(x,y) &= \frac{1}{8}(-5+6x+7y)(\chi_{\Omega_1}+\chi_{\Omega_2}+\chi_{\Omega_3}) &+ \frac{1}{8}(-3-11x+34y)\chi_{\Omega_0}, \\ w_{18}(x,y) &= \frac{1}{8}(-3+5x+2y)(\chi_{\Omega_1}+\chi_{\Omega_2}+\chi_{\Omega_3}) &+ \frac{1}{8}(-3-11x+34y)\chi_{\Omega_0} \end{split}$$

The pictures of w_{0j} , $j \in \mathbb{Z}_3$ and w_{1k} , $k \in \mathbb{Z}_9$ are shown in Fig. 3.2 through Fig. 3.13.

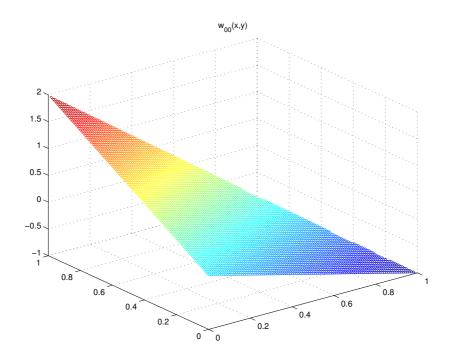


Figure 3.2: Scaling function w_{00}

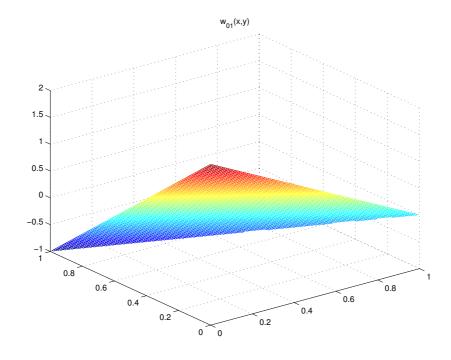


Figure 3.3: Scaling function w_{01}

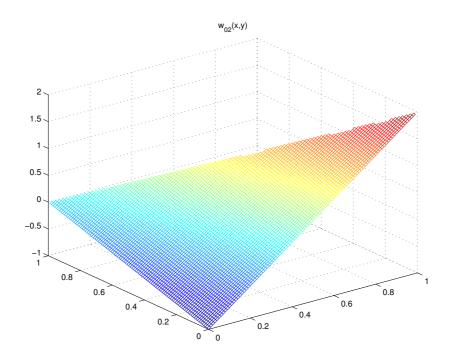


Figure 3.4: Scaling function w_{02}

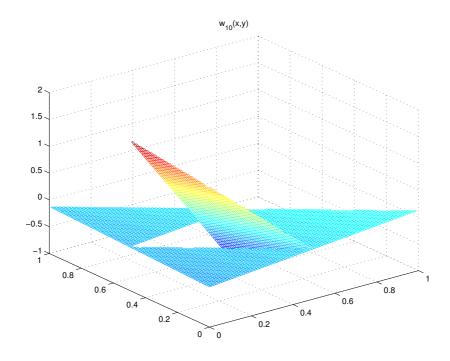


Figure 3.5: Wavelet function w_{10}

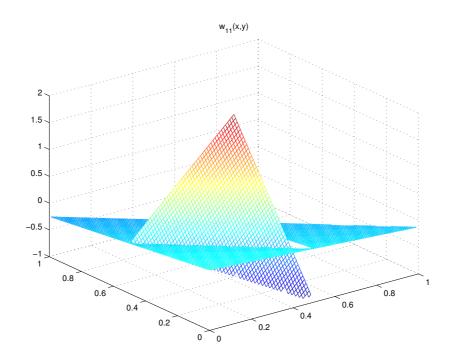


Figure 3.6: Wavelet function w_{11}

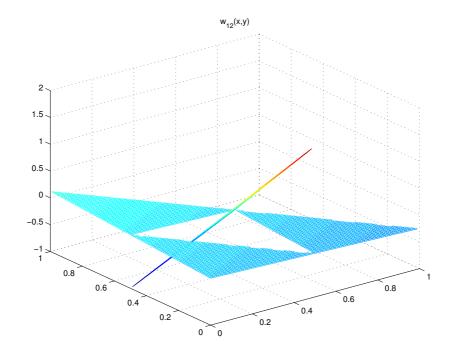


Figure 3.7: Wavelet function w_{12}

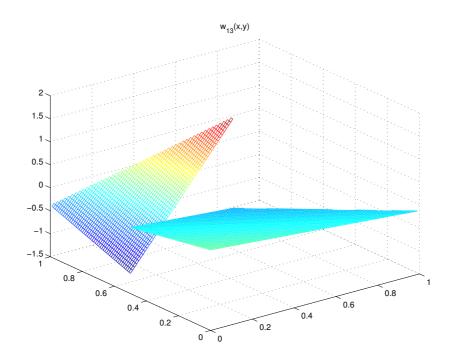


Figure 3.8: Wavelet function w_{13}

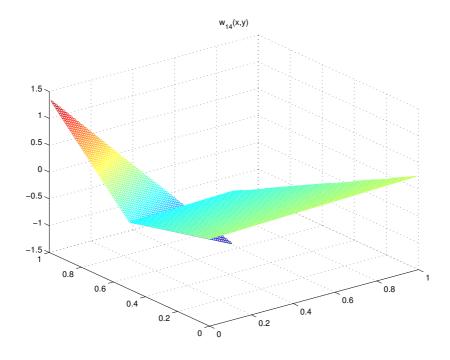


Figure 3.9: Wavelet function w_{14}

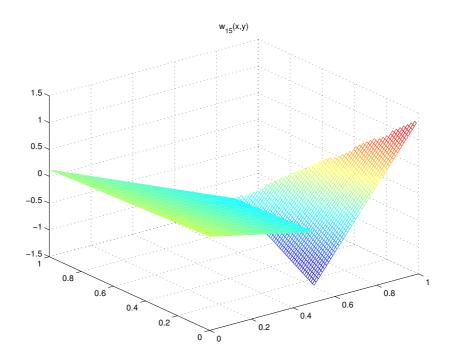


Figure 3.10: Wavelet function w_{15}

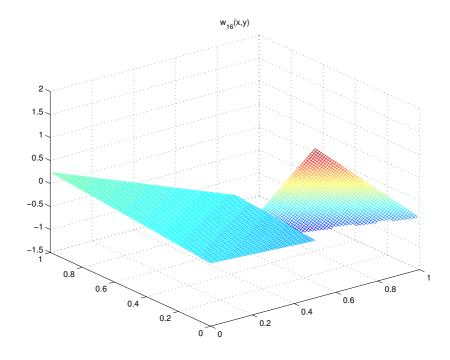


Figure 3.11: Wavelet function w_{16}

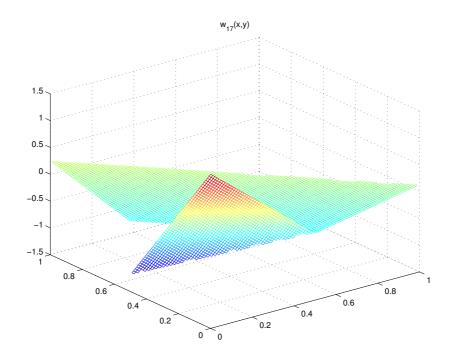


Figure 3.12: Wavelet function w_{17}

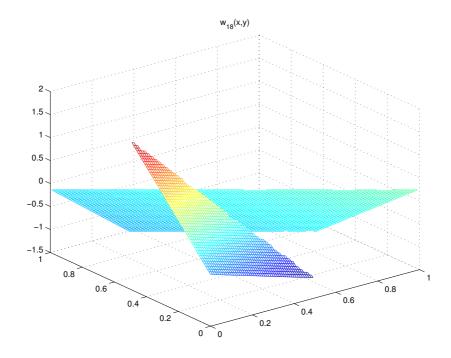


Figure 3.13: Wavelet function w_{18}

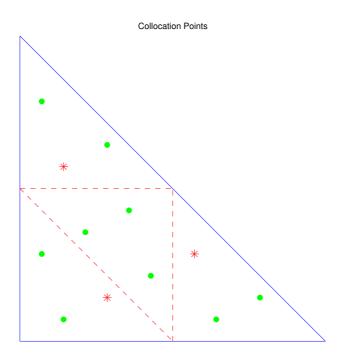


Figure 3.14: Collocation points of L_1

3.3 Initial Collocation Functionals in L_1

We define the set of initial collocation functionals at level 0 by

$$L_0 := \{ \ell_{0j} = \delta_{t_{0j}}, \ j \in \mathbb{Z}_3 \}.$$

To find $L_1 := \{\ell_{1j} : j \in \mathbb{Z}_9\}$, introduce a vector

$$\mathbf{L} = [\ell_{10}, \ell_{11}, \ell_{12}, \ell_{13}, \ell_{14}, \ell_{15}, \ell_{16}, \ell_{17}, \ell_{18}]^T,$$
(3.27)

where, for $k \in \mathbb{Z}_9$,

$$\ell_{1k} = \sum_{m \in \mathbb{Z}_{12}} d_{km} \delta_{t_{1m}}, \quad t_{1m} = \phi_i t_{0j}, \quad m = 3i + j, \quad i \in \mathbb{Z}_4, \ j \in \mathbb{Z}_3.$$
(3.28)

The 12 collocation points t_{1m} , $m \in \mathbb{Z}_{12}$ are plotted in Fig. 3.14 (The three points marked by asterisk are the three initial collocation points t_{0j} , $j \in \mathbb{Z}_3$). Let matrix $\mathbf{C}' \in M_{9,12}$ with

$$\mathbf{C}' = [d_{km} : k \in \mathbb{Z}_9, \ m \in \mathbb{Z}_{12}].$$
(3.29)

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Define

$$\delta_{\mathbf{t}} = [\delta_{t_{1,0}}, \delta_{t_{1,1}}, \delta_{t_{1,2}}, \delta_{t_{1,3}}, \delta_{t_{1,4}}, \delta_{t_{1,5}}, \delta_{t_{1,6}}, \delta_{t_{1,7}}, \delta_{t_{1,8}}, \delta_{t_{1,9}}, \delta_{t_{1,10}}, \delta_{t_{1,11}}]^T
= [\delta_{(\frac{1}{14}, \frac{2}{7})}, \delta_{(\frac{1}{7}, \frac{1}{14})}, \delta_{(\frac{2}{7}, \frac{1}{7})}, \delta_{(\frac{4}{7}, \frac{2}{7})}, \delta_{(\frac{9}{14}, \frac{1}{14})}, \delta_{(\frac{11}{14}, \frac{1}{7})}, \delta_{(\frac{1}{14}, \frac{11}{14})}, \delta_{(\frac{1}{7}, \frac{4}{7})}, \delta_{(\frac{2}{7}, \frac{9}{14})}, \qquad (3.30)
\qquad \delta_{(\frac{3}{7}, \frac{3}{14})}, \delta_{(\frac{5}{14}, \frac{3}{7})}, \delta_{(\frac{3}{14}, \frac{5}{14})}]^T.$$

Then we have

$$\mathbf{L} = \mathbf{C}' \delta_{\mathbf{t}}.\tag{3.31}$$

Again, to construct ℓ_{1k} , $k \in \mathbb{Z}_9$ having vanishing moment, require

where $k, j \in \mathbb{Z}_9$. (3.32) leads to nine 12×12 linear systems. After solving these nine 12×12 linear systems, we find that

$$\mathbf{C}' = \begin{bmatrix} 0 & 0 & -\frac{1}{2} & 0 & 0 & 0 & 0 & -\frac{1}{2} & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & -\frac{1}{2} & 0 & 0 & 0 & -\frac{1}{2} & 0 & 0 & 1 & 0 \\ 0 & 0 & -\frac{1}{2} & -\frac{1}{2} & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & -\frac{1}{2} & 0 & 0 & 0 & -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 1 & -\frac{3}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -\frac{3}{2} & 0 & 1 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & -\frac{1}{2} & -1 & 1 & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 1 & -\frac{3}{2} & \frac{1}{2} & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & -1 & \frac{1}{2} & 0 & 0 & 0 & -\frac{1}{2} & 0 & 0 & 0 & 0 \end{bmatrix}.$$

$$(3.33)$$

To this point we have obtained desired w_{1j} and ℓ_{1j} , $j \in \mathbb{Z}_9$ having vanishing moment. To find the remaining multiscale wavelets w_{ij} and functionals ℓ_{ij} , for $i \ge 2$, $j \in w(i)$, simply apply (2.40) and (2.50).

3.4 Properties of the Wavelet Basis Functions and the Collocation Functionals

We examine properties of the wavelet basis functions and the collocation functionals constructed above. These properties are crucial for developing fast wavelet collocation methods. Next several lemmas tells us these properties.

Lemma 3.1 The constructed basis function w_{ij} , $i \in \mathbb{N}_0$, $j \in \mathbb{Z}_{w(i)}$ are locally supported and their supports are shrinking as level *i* increases.

Proof: for i > 1, the support of w_{ij} is contained in $S_{ij} = \phi_{\mathbf{e}}(E) = E_{i-1,\mu(\mathbf{e})}$, where $j = 9\mu(\mathbf{e}) + l$, $l \in \mathbb{Z}_9$. Note $\phi_{\mathbf{e}}$ is a contractive mapping by the contractivity of ϕ_i , $i \in \mathbb{Z}_4$. Hence, S_{ij} is shrinking as *i* increases.

Lemma 3.2 For any $i, i' \in \mathbb{N}_0$,

$$\langle \ell_{i'j'}, w_{ij} \rangle = \delta_{ii'} \delta_{jj'}, \quad (i,j), (i',j') \in \mathbb{U}, \ i \le i', \tag{3.34}$$

and

$$\sum_{j \in \mathbb{Z}_{w(i)}} |\langle \ell_{i'j'}, w_{ij} \rangle| \le \frac{81}{28}, \quad (i,j), (i',j') \in \mathbb{U}, \ i > i'.$$
(3.35)

Further we have

$$\gamma < \mu^{k/2} - 1, \tag{3.36}$$

where, γ is defined in (3.37), and k is the total degree of the pieced polynomial spaces $\mathbb{W}_n, n \in \mathbb{N}_0$.

Proof: (3.34) is proved in Lemma 4.3. Let $\mathbf{C}_1 = [c_{ij}, i \in \mathbb{Z}_9, j \in \mathbb{Z}_3]$, where c_{ij} are elements of \mathbf{C} (see (3.22)). Note $\|\mathbf{C}_1\|_1 = \frac{27}{28}$, and $\|\mathbf{C}'\|_{\infty} = 3$. Set

$$\gamma = \max\{\|\mathbf{C}_1\|_1, \|\mathbf{C}'\|_{\infty} \|\mathbf{C}_1\|_1\} = \frac{81}{28}.$$
(3.37)

By Lemma 5.2 of [CMX3], which says $\sum_{j \in \mathbb{Z}_{w(i)}} |\langle \ell_{i'j'}, w_{ij} \rangle| \leq \gamma$, we have (3.35). Direct computation shows (3.36) with k = 2.

Lemma 3.3 The constructed wavelets w_{ij} and collocation functionals ℓ_{ij} for $i \in \mathbb{N}_0$, $j \in \mathbb{Z}_{w(i)}$ have vanishing moment of degree 2.

Proof: From (3.32), we easily see

$$\langle \ell_{i'j'}, w_{0j} \rangle = \langle \ell_{1l'}, w_{0j} \circ \phi_{\mathbf{e}'} \rangle = 0, \quad j \in \mathbb{Z}_3.$$
(3.38)

Now consider $\mathcal{T}_{\mathbf{e}}$ as an operator from $L^2(E)$ to $L^2(E)$ and let $\mathcal{T}_{\mathbf{e}}^*$ be the adjoint of $\mathcal{T}_{\mathbf{e}}$. It is easy to see that for $y \in L^2(E)$

$$\mathcal{T}_{\mathbf{e}}^* y = J_{\phi_{\mathbf{e}}} y \circ \phi_{\mathbf{e}},\tag{3.39}$$

where $J_{\phi_{\mathbf{e}}}$ is the Jacobi of mapping $\phi_{\mathbf{e}}$. Therefore,

$$(w_{ij}, w_{0j'}) = (\mathcal{T}_{\mathbf{e}} w_{1l}, w_{0j'}) = (w_{1l}, \mathcal{T}_{\mathbf{e}}^* w_{0j'}) = 0.$$

The last equality holds because $\mathcal{T}_{\mathbf{e}}^* w_{0j'}$ is a polynomial of total degree $\leq k-1 = 1$, with k = 2and w_{1l} satisfies (3.17).

Thus for any polynomials $p \in \pi_2$,

$$\langle \ell_{ij}, p \rangle = 0, \quad (w_{ij}, p) = 0, \quad (i, j) \in \mathbb{U}, \ i \ge 1,$$
(3.40)

where π_k is the space of polynomials total degree less than k.

Lemma 3.4 Both basis functions and collocation functionals are uniformly bounded.

Proof: For $(i, j) \in \mathbb{U}$, $i \geq 2$, $j = \mu(\mathbf{e})r + l$, $l \in \mathbb{Z}_r$,

$$|\langle \ell_{ij}, v \rangle| = |\langle \ell_{1l}, v \circ \phi_{\mathbf{e}} \rangle| \le \|\mathbf{C}'\|_{\infty} \|v\|_{\infty}$$

and

$$\|w_{ij}\|_{\infty} \leq \|w_{1l} \circ \phi_{\mathbf{e}}^{-1} \chi_{\phi_{\mathbf{e}}(E)}\|_{\infty} \leq \|\mathbf{C}\|_{\infty} \max_{j \in \mathbb{Z}_{12}} \|\psi_j\|_{\infty}.$$

Lemma 3.5 The dimension of spaces \mathbb{F}_n and \mathbb{W}_n grows exponentially in n and the diameters d_n decay exponentially in n too. where

$$d_n := max\{d(S_{nm}) : m \in \mathbb{Z}_{w(n)}\}, \quad n \in \mathbb{N}_0$$

Proof: We have dim $\mathbb{W}_0 = 3$, dim $\mathbb{F}_n = 3.4^n$, $n \in \mathbb{N}_0$, and dim $\mathbb{W}_n = 9.4^{n-1}$, $n \ge 1$, $n \in \mathbb{Z}$. For the diameters we have $d_0 = \sqrt{2}$ and $d_n = \sqrt{2}\frac{1}{2^{n-1}} = 2^{3/2-n}$.

Lemma 3.6 There exists constant θ_2 and θ_3 such that for all $n \in \mathbb{N}_0$ and v having form

$$v := \sum_{(i,j)\in\mathbb{U}_n} v_{ij} w_{ij} \tag{3.41}$$

 $\theta_2 \|\mathbf{v}\|_{\infty} \le \|v\|_{\infty} \le \theta_3 (n+1) \|\mathbf{E}_n \mathbf{v}\|_{\infty}$ (3.42)

where $\mathbf{v} := [v_{ij} : (i, j) \in \mathbb{U}_n]^T$.

Proof: The first inequality follows from PROPOSITION 5.6 of [CMX3]. For the second inequality, we provide here a slight different proof. Define $\{\zeta_{ij} : (i, j) \in \mathbb{U}\}$ by:

$$\zeta_{0j} := w_{0j}, \quad j \in \mathbb{Z}_3. \tag{3.43}$$

For $j \in \mathbb{Z}_9$, find vector $\mathbf{c}''_{js} := [c''_{js} : s \in \mathbb{Z}_{12}]$ such that

$$\zeta_{1j} := \sum_{s \in \mathbb{Z}_{12}} c_{js}'' \psi_s \tag{3.44}$$

satisfies

$$\langle \ell_{0j'}, \zeta_{1j} \rangle = 0, \quad j' \in \mathbb{Z}_3, \tag{3.45}$$

and

$$\langle \ell_{1j'}, \zeta_{1j} \rangle = \delta_{jj'}, \quad j' \in \mathbb{Z}_9.$$
 (3.46)

Further define

$$\zeta_{ij} := \mathcal{T}_{\mathbf{e}} \zeta_{1l}. \tag{3.47}$$

A direct computation shows that

$$\langle \ell_{i'j'}, \zeta_{ij} \rangle = \delta_{ii'} \delta_{jj'}, \quad (i,j), (i',j') \in \mathbb{U}.$$
(3.48)

[CMX3] proves in its PROPOSITION 5.6 that there exists a constant θ_3 , for all $i \in \mathbb{N}_0$,

$$\sup_{t \in \Omega} \sum_{j \in \mathbb{Z}_{w(i)}} |\zeta_{ij}(t)| \le \theta_3$$
(3.49)

For $(i, j) \in \mathbb{U}_n$, $\{w_{ij}\}$ and $\{\zeta_{ij}\}$ are two different bases of \mathbb{F}_n . So we have

$$v = \sum_{(i,j)\in\mathbb{U}_n} v_{ij} w_{ij} = \sum_{(i,j)\in\mathbb{U}_n} \tilde{v}_{ij} \zeta_{ij}.$$
(3.50)

Observe that $\mathbf{E}_n \mathbf{v} = \tilde{\mathbf{v}}$, where $\tilde{\mathbf{v}} := [\tilde{v}_{ij} : (i, j) \in \mathbb{U}_n]^T$. Now

$$\|v\|_{\infty} = \|\sum_{(i,j)\in\mathbb{U}_n} \tilde{v}_{ij}\zeta_{ij}\|_{\infty} \leq \|\tilde{\mathbf{v}}_{\infty}\sum_{(i,j)\in\mathbb{U}_n} \zeta_{ij}\|_{\infty} \leq \tilde{\mathbf{v}}_{\infty}\|\sum_{i\in\mathbb{Z}_{n+1}} \sum_{j\in\mathbb{Z}_{w(i)}} \zeta_{ij}\|_{\infty} \leq \theta_3(n+1)\|\mathbf{E}_n\mathbf{v}\|_{\infty}.$$
(3.51)

This completes the proof.

Lemma 3.7 Let \mathcal{P}_n be the projection operator from \mathbb{X} onto \mathbb{F}_n defined by the requirement that

$$\langle \ell_{ij}, \mathcal{P}_n x \rangle = \langle \ell_{ij}, x \rangle, \ (i, j) \in \mathbb{U}_n.$$
 (3.52)

For $v \in L^{\infty}(E)$ we set

$$\mathcal{P}_n v = \sum_{(i,j) \in \mathbb{U}_n} v_{ij} w_{ij}$$

then

$$\lim_{n \to \infty} \|\mathcal{P}_n x - x\|_{\infty} = 0.$$
(3.53)

Proof: Since \mathbb{F}_n is a space of piecewise polynomials, the operator \mathcal{P}_n converges pointwise to the identity operator \mathcal{I} in $L^{\infty}(E)$ as $n \to \infty$ by results from [AGS]. In other words, (3.53) holds.

Remark: For $x \in \mathbb{X}$, assume $\mathcal{P}_n x = \sum_{(i,j) \in \mathbb{U}_n} v_{ij} w_{ij}$. If we define a vector $\mathbf{L}_n := \langle \ell_{ij}, x \rangle, (i,j) \in \mathbb{U}_n$, then

$$\mathbf{E}_n \mathbf{v} = \mathbf{L}_n \quad \text{or } \mathbf{v} = \mathbf{E}_n^{-1} \mathbf{L}_n. \tag{3.54}$$

Lemma 3.8 There exists a positive constant c such that for $u \in W^{k,\infty}(E)$

$$\operatorname{dist}(u, \mathbb{F}_n) \le c\mu^{-kn/2} \|u\|_{k,\infty}.$$
(3.55)

where for $\alpha := [\alpha_i \in \mathbb{N}_0 : i \in \mathbb{Z}_2], \ |\alpha| := \sum_{i \in \mathbb{Z}_2} \alpha_i$

$$||u||_{k,\infty} := \max\{||D^{\alpha}u||_{\infty} : |\alpha| \le k\},\$$

and μ is the number of contractive mappings ϕ_i , $i \in \mathbb{Z}_{\mu}$, as in (3.1), $\mu = 4$.

Proof: Since \mathbb{F}_n are the spaces of piecewise polynomials of total degree $\leq k - 1$ and our assumption that E is a star-shaped set, then there exists a positive constant c such that

$$dist(u, \pi_k) \le c(d(E))^k ||u||_{k,\infty}.$$
 (3.56)

This together with (2.9) competes the proof.

3.5 Reconstruction of Functions

Let E be the unit right triangle, $P_0 = (0,0)$, $P_1 = (1,0)$. The vertices of the domain $\hat{E} := \phi_{\mathbf{e}}(E)$ corresponding to P_0, P_1 are denoted by $\hat{P}_i, i \in \mathbb{Z}_2$, namely, $\hat{P}_i = \phi_{\mathbf{e}}(P_i), i \in \mathbb{Z}_2$. Further $\hat{P}_i = (\hat{x}_i, \hat{y}_i), i \in \mathbb{Z}_2$. Observe that

$$\hat{E} := \phi_{\mathbf{e}}(E) = \begin{cases} \{(s,t) : \hat{x}_0 \le s \le \hat{x}_1, \hat{y}_0 \le t \le -s + (\hat{x}_1 + \hat{y}_1)\} & \text{if } \hat{x}_0 \le \hat{x}_1, \\ \{(s,t) : \hat{x}_1 \le s \le \hat{x}_0, -s + (\hat{x}_1 + \hat{y}_1) \le t \le \hat{y}_0\} & \text{if } \hat{x}_0 \ge \hat{x}_1. \end{cases}$$
(3.57)

Let u be the solution of (2.2) in \mathbb{X} , u_n be the approximation of u sought by wavelet collocation scheme in \mathbb{F}_n . We have known from (2.16) that

$$u_n(x) := \sum_{(i,j) \in \mathbb{U}_n} u_{ij} w_{ij}(x)$$
 (3.58)

where, $\mathbf{u}_n = [u_{ij} : (i, j) \in \mathbb{U}_n]$ is found by previously described wavelet collocation scheme. Since each wavelet w_{ij} has compact support, many terms in the summation of (3.58) are zeros. We need to determine if $w_{ij}(x)$ is vanished or not. If not, evaluate the value $w_{ij}(x)$. By (2.40), (2.39) and (2.34) we have

$$w_{ij}(x) = \mathcal{T}_{\mathbf{e}} w_{1l}(x) = w_{1l}(\phi_{\mathbf{e}}^{-1}(x))\chi_{\phi_{\mathbf{e}}(E)}(x), \qquad (3.59)$$

where $j = \mu(\mathbf{e})r + l$, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $l \in \mathbb{Z}_r$, r = 9, $\mu = 4$.

We describe a *fast* algorithm in the following to compute $u_n(x)$. Assume x is given and $\mathbf{u}_n = [u_{ij} : (i, j) \in \mathbb{U}_n]$ is found.

Algorithm:

1. Compute $u_n(x) := \sum_{(i,j) \in \mathbb{U}_1} u_{ij} w_{ij}(x);$

```
2. for i = 2; i ≤ n; step 1, loop
for j = 0; j < μ<sup>i-1</sup>; step 1, loop
find e ∈ Z<sup>i-1</sup><sub>μ</sub>, such that j = μ(e);
if x ∈ φ<sub>e</sub>(E), compute the r evaluations w<sub>1l</sub>(φ<sup>-1</sup><sub>e</sub>(x)), l ∈ Z<sub>r</sub>, and sum them up.
Then jump out of the j loop, go to the next i loop;
if x ∉ φ<sub>e</sub>(E), continue next j loop;
end loop;
```

Note in (3.58) for each level $i \ge 1$ and fixed $x \in E$, only r terms in the summation are nonzero, namely, only r terms of $\chi_{\phi_{\mathbf{e}}(E)}(x) = 1$. Also note to compute $w_{ij}(x)$, instead of directly using the expression of w_{ij} , we always go back to the first level w_{1l} by the recursive formula (3.59).

Chapter 4

Computation of Discrete Systems and Their Solutions

We now turn our attention to compute matrices \mathbf{E}_n and \mathbf{K}_n of the discretized system (2.18). Their elements are expressed by (2.19) and (2.20) respectively. To compute these entries, we shall need to find ℓ_{ij} , w_{ij} , for $i \ge 2$, $j \in \mathbb{Z}_{w(i)}$. Although (2.40) and (2.50) give expressions to find w_{ij} , ℓ_{ij} , $i \ge 2$, $j \in \mathbb{Z}_{w(i)}$ respectively, they are not convenient for actual computing. More direct relations to wavelets and collocation functionals of level 1 are needed to easily compute w_{ij} , ℓ_{ij} , for $i \ge 2$, $j \in \mathbb{Z}_{w(i)}$. Again we set $\mu = 4$ and r = 9. By (2.50) and (3.28) it is easy to see that

$$\ell_{ij} = \sum_{m \in \mathbb{Z}_{12}} d_{lm} \delta_{t_{im}}, \quad i \ge 2, \ j \in \mathbb{Z}_{w(i)}, \tag{4.1}$$

where, $l = j \mod r$; d_{lm} , $l \in \mathbb{Z}_9$, $m \in \mathbb{Z}_{12}$ are defined in (3.29), and

$$t_{im} = \phi_{\mathbf{e}}(t_{1m}), \quad m \in \mathbb{Z}_{12}, \tag{4.2}$$

with $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$. For wavelets $w_{ij}, i \geq 2, j \in \mathbb{Z}_{w(i)}$, in view of (2.40) we have

$$w_{ij} = w_{1l} \circ \phi_{\mathbf{e}}^{-1}(t) \chi_{\phi_{\mathbf{e}}(E)}(t), \qquad (4.3)$$

where $j = \mu(\mathbf{e})r + l$, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, and $l \in \mathbb{Z}_r$.

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The Matrix \mathbf{E}_n 4.1

Denote the element of \mathbf{E}_n by $E_{i'j',ij}$, where $(i,j), (i',j') \in \mathbb{U}_n$. By (2.20), (4.1) and (4.3), we obtain

$$E_{i'j',ij} = \langle \ell_{i'j'}, w_{ij} \rangle$$

$$= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \langle \delta_{t_{i'm}}, w_{ij} \rangle$$

$$= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \langle \delta_{\phi_{\mathbf{e}'}(t_{1m})}, w_{ij} \rangle$$

$$= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} w_{1l} (\phi_{\mathbf{e}}^{-1} \circ \phi_{\mathbf{e}'}(t_{1m})) \chi_{\phi_{\mathbf{e}}(E)} (\phi_{\mathbf{e}'}(t_{1m})) .$$

$$(4.4)$$

where, $j = \mu(\mathbf{e})r + l$, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $l \in \mathbb{Z}_r$ and $j' = \mu(\mathbf{e}')r + l'$, $\mathbf{e}' \in \mathbb{Z}_{\mu}^{i'-1}$, $l' \in \mathbb{Z}_r$. Define

$$\mathbf{E}_{i'i} := [E_{i'j',ij}: j' \in \mathbb{Z}_{w(i')}, j \in \mathbb{Z}_{w(i)}]$$

as the submatrix of \mathbf{E}_n ($\mathbf{K}_{i'i}$ is similarly defined). We further partition $\mathbf{E}_{i'i}$ (also $\mathbf{K}_{i'i}$) into blocks. Each block of $\mathbf{E}_{i'i}$ corresponds to a pair of supports $(\hat{S}_{i'j'}, S_{ij})$ which are supports of the pair $(\ell_{i'j'}, w_{ij})$ respectively.

Definition 4.1 A set of elements in the submatrix $\mathbf{E}_{i'i}$ (also in $\mathbf{K}_{i'i}$) is called a block if they correspond to the same pair of supports $(\hat{S}_{i'j'}, S_{ij})$ associated with the pair $(\ell_{i'j'}, w_{ij})$.

Denote by the pair (S(i'), S(i)) the size of blocks in submatrix $\mathbf{E}_{i'i}$ (also in $\mathbf{K}_{i'i}$), with

$$S(k) = \begin{cases} m, & k = 0 \\ r, & otherwise, \end{cases}$$
(4.5)

where, $m = \dim \mathbb{F}_0$. In our case m = 3. By Definition 4.1, when i' = 0 and i = 0, the block has size $m \times m$, and for i', i > 0, each block has size $r \times r$. Therefore $\mathbf{E}_{i'i}$ (also $\mathbf{K}_{i'i}$) has $R(i') = \frac{w(i')}{S(i')}$ rows of blocks and $C(i) = \frac{w(i)}{S(i)}$ columns of blocks. For brevity, we write $\mathbf{E}_{i'i}$ as

$$\mathbf{E}_{i'i} = [b_{jk}]_{j \in \mathbb{Z}_{R(i')}, k \in \mathbb{Z}_{C(i)}}.$$
(4.6)

Following when describing the computation of matrix $\mathbf{E}_{i'i}$ (also $\mathbf{K}_{i'i}$), we treat each b_{jk} as a matrix element.

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Definition 4.2 For a matrix $\mathbf{A} = [a_{jk}]_{m \times n}$, where m may not be equal to n but one dividing the other. We call $\theta = \frac{max\{m,n\}}{min\{m,n\}}$ the shape index of \mathbf{A} . The main diagonal line of matrix \mathbf{A} is defined as the set

$$\{a_{jk}: j\theta \le k < (j+1)\theta\} \quad \text{if } m \le n, \tag{4.7}$$

or

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$$\{a_{jk}: k\theta \le j < (k+1)\theta\} \quad \text{if } m > n.$$

$$(4.8)$$

For a given positive integer c the c-th lower diagonal is defined as the set

$$\{a_{jk}: (j-c)\theta \le k < (j-c+1)\theta\}$$
 if $m \le n$, (4.9)

or

$$\{a_{jk}: (k+c)\theta \le j < (k+c+1)\theta\} \text{ if } m > n.$$
(4.10)

The c-th upper diagonal is defined as the set

$$\{a_{jk}: (j+c)\theta \le k < (j+c+1)\theta\}$$
 if $m \le n$, (4.11)

or

$$\{a_{jk}: (k-c)\theta \le j < (k-c+1)\theta\} \text{ if } m > n.$$
(4.12)

Note, when c = 0 we obtain the main diagonal in any case.

Next Lemma shows that \mathbf{E}_n is upper triangular with diagonal blocks equal to identities of corresponding sizes. Moreover, the upper half of \mathbf{E}_n is sparse already.

Lemma 4.3 For matrix $\mathbf{E}_n := [E_{i'j',ij}]$, where $i', i \in \mathbb{Z}_{n+1}, j \in \mathbb{Z}_{(i)}$, and $j' \in \mathbb{Z}_{(i')}$.

- 1. $E_{i'j',ij} = 0$, for i' > i;
- 2. $E_{i'j',ij} = \delta_{i'i}\delta_{jj'}$, for i' = i;
- 3. and, if i' < i, in each submatrix $\mathbf{E}_{i'i}$ with shape parameter $\theta_{i'i}$, only elements on the main diagonal line are possibly nonzero.

Proof: For $(i, j) \in \mathbb{U}_n$, with $i \geq 2$, there exists a unique pair of $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$ and $l \in \mathbb{Z}_r$ such that $j = \mu(\mathbf{e})r + l$ and $w_{ij} = \mathcal{T}_{\mathbf{e}}w_{1l}$. Likewise, for $(i', j') \in \mathbb{U}_n$, with $i' \geq 2$, there exists a unique pair of $\mathbf{e}' \in \mathbb{Z}_{\mu}^{i'-1}$ and $l' \in \mathbb{Z}_r$ such that $j' = \mu(\mathbf{e}')r + l'$ and $\ell_{i'j'} = \mathcal{L}_{\mathbf{e}'}\ell_{1l'}$. When i = i', it follows from (2.49) and (2.45) that

$$\langle \ell_{i'j'}, w_{ij} \rangle = \langle \mathcal{L}_{\mathbf{e}'} \ell_{1l'}, \mathcal{T}_{\mathbf{e}} w_{1l} \rangle = \langle \ell_{1l'}, w_{1l} \rangle \delta_{\mathbf{e}'\mathbf{e}} = \delta_{l'l} \delta_{\mathbf{e}\mathbf{e}'} = \delta_{j'j},$$

since $\mathbf{e} = \mathbf{e}'$. When i < i', let $\mathbf{e}'_1 = (e'_0, \dots, e'_{i-2}), \ \mathbf{e}'_2 = (e'_{i-1}, \dots, e'_{i'-2})$, then

$$\langle \ell_{i'j'}, w_{ij} \rangle = \langle \mathcal{L}_{\mathbf{e}_2'} \ell_{1l'}, w_{1l} \rangle \delta_{\mathbf{e}_1'\mathbf{e}} = \langle \ell_{1l'}, w_{1l} \circ \phi_{\mathbf{e}_2'} \rangle \delta_{\mathbf{e}_1'\mathbf{e}}.$$

Since $\phi_{\mathbf{e}'_2} : E \to \phi_{\mathbf{e}'_2}(E)$ is an affine mapping, $w_{1l} \circ \phi_{\mathbf{e}'_2}$ is a polynomial of total degree $\leq k-1$ in \mathbb{F}_0 . By using (2.44), we have that

$$\langle \ell_{i'j'}, w_{ij} \rangle = 0, \quad (i,j), (i',j') \in \mathbb{U}_n, i < i'.$$

when i > i', let $\mathbf{e} = (\mathbf{e}_1, \mathbf{e}_2)$, where $\mathbf{e}_1 = (e_0, \dots, e_{i'-2})$ and $\mathbf{e}_2 = (e_{i'-1}, \dots, e_{i-2})$, and

$$\langle \ell_{i'j'}, w_{ij} \rangle = \langle \mathcal{L}_{\mathbf{e}'} \ell_{1l'}, \mathcal{T}_{\mathbf{e}} w_{1l} \rangle = \langle \ell_{1l'}, \mathcal{T}_{\mathbf{e}_2} w_{1l} \rangle \delta_{\mathbf{e}'\mathbf{e}_1}$$
(4.13)

For the elements on the main diagonal line of submatrix $\mathbf{E}_{i'i}$, $\mathbf{e}' = \mathbf{e}_1$, and for those elements off the main diagonal line $\mathbf{e}' \neq \mathbf{e}_1$. This completes the proof.

By (4.4) and (4.13), t for the elements on the main diagonal line of $\mathbf{E}_{i'i}$,

$$E_{i'j',ij} = \sum_{m \in \mathbb{Z}_{12}} d_{l'm} w_{1l} \circ \phi_{\mathbf{e}_2}^{-1}(t_{1m}) \chi_{\phi_{\mathbf{e}_2}(E)}(t_{1m}).$$
(4.14)

(4.14) actually reduces the computation of the elements of \mathbf{E}_n compared to (4.4). Fig. 4.1, Fig. 4.2 and Fig. 4.3 demonstrate computed full matrices \mathbf{E}_n after discretizing the integral equation (6.1) with n = 2, 3, 4 respectively. Clearly \mathbf{E}_n is upper-triangular and sparse.

4.2 The Matrix \mathbf{K}_n

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In view of (2.19), all the elements of \mathbf{K}_n involve integrals of type $\int_{\Omega} K(x, y) w_{ij}(y) dy$, where, $x \in \mathbb{R}^2$. Following theorem presents a perspective to compute this type of integrals.

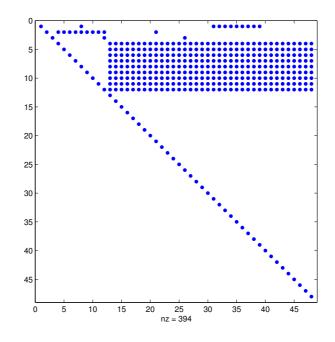


Figure 4.1: Matrix \mathbf{E}_2

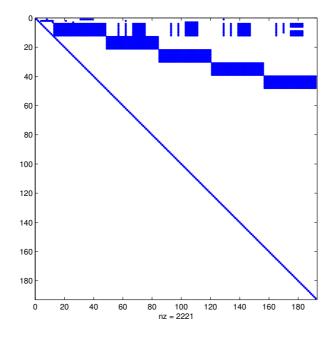


Figure 4.2: Matrix \mathbf{E}_3

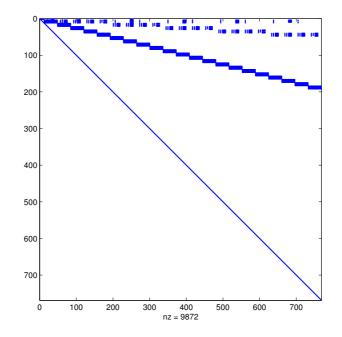


Figure 4.3: Matrix \mathbf{E}_4

Theorem 4.4

$$\int_{\Omega} K(x,y)w_{ij}(y)dy = J_{\phi_{\mathbf{e}}} \int_{\Omega} K(x,\phi_{\mathbf{e}}(y))w_{1l}(y)dy, \qquad (4.15)$$

where $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $j = \mu(\mathbf{e}) + l$, $l \in \mathbb{Z}_r$, $J_{\phi_{\mathbf{e}}}$ is the Jacobian of mapping $\phi_{\mathbf{e}}$. Namely,

$$J_{\phi_{\mathbf{e}}} = J_{\phi_{e_0}} \cdot J_{\phi_{e_1}} \cdots J_{\phi_{e_{i-3}}} \cdot J_{\phi_{e_{i-2}}}, \tag{4.16}$$

and w_{ij} is obtained by (2.40) for $i \ge 2, i \in \mathbb{Z}$.

Proof: We shall prove Theorem 4.4 by induction on i. Let i = 2, $\mathbf{e} \in \mathbb{Z}^1_{\mu}$. Assume $\mathbf{e} = [e_0]$. Then

$$\int_{\Omega} K(x, y) w_{2j}(y) dy$$

$$= \int_{\Omega} K(x, y) \mathcal{T}_{\mathbf{e}} w_{1l}(y) dy$$

$$= \int_{\Omega} K(x, y) \mathcal{T}_{e_0} w_{1l}(y) dy$$

$$= \int_{\Omega} K(x, y) w_{1l}(\phi_{e_0}^{-1}(y)) dy$$

$$= J_{\phi_{e_0}} \int_{\Omega} K(x, \phi_{e_0}(y)) w_{1l}(y) dy.$$

The last equality is obtained by changing variable. Hence (4.15) holds for the case i = 2. Assume (4.15) holds for i < n. for i = n, $\mathbf{e} \in \mathbb{Z}_{\mu}^{n-1}$. Denote $\mathbf{e} = [e_0 e_1 \cdots e_{n-2}]$. We have

$$\int_{\Omega} K(x, y) w_{nj}(y) dy$$

$$= \int_{\Omega} K(x, y) \mathcal{T}_{\mathbf{e}} w_{1l}(y) dy$$

$$= \int_{\Omega} K(x, y) \mathcal{T}_{e_0, e_1, \dots, e_{n-2}} w_{1l}(y) dy$$

$$= \int_{\Omega} K(x, y) \mathcal{T}_{e_0} \mathcal{T}_{e_1, \dots, e_{n-2}} w_{1l}(y) dy$$

$$= J_{\phi_{e_0}} \int_{\Omega} K(x, \phi_{e_0}(y)) \mathcal{T}_{e_1, \dots, e_{n-2}} w_{1l}(y) dy$$

$$= J_{\phi_{\mathbf{e}}} \int_{\Omega} K(x, \phi_{\mathbf{e}}(y)) w_{1l}(y) dy.$$

The last equality obtained by the induction assumption.

Theorem 4.4 facilitates us a convenient way to compute the integral on the left side of (4.15). Note w_{ij} has compact support $\phi_{\mathbf{e}}(E)$. But by Theorem 4.4 we don't need to find the explicit expression for w_{ij} and its support $\phi_{\mathbf{e}}(E)$.

We now look at how to compute elements of \mathbf{K}_n . Denote elements of \mathbf{K}_n by $K_{i'j',ij}$, where, $(i,j), (i',j') \in \mathbb{U}_n$. For $i, i \geq 2$, let $j = \mu(\mathbf{e})r + l$, and $j' = \mu(\mathbf{e}')r + l'$, where, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $\mathbf{e}' \in \mathbb{Z}_{\mu}^{i'-1}$, and $l, l' \in \mathbb{Z}_r$. By (2.19), (4.1), (4.2) and Theorem 4.4

$$K_{i'j',ij} = \langle \ell_{i'j'}, \mathcal{K}w_{ij} \rangle$$

$$= \int_{\Omega} \langle \ell_{i'j'}, K(\cdot, y) \rangle w_{ij}(y) dy$$

$$= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\Omega} K(\phi_{\mathbf{e}'}(t_{1m}), y) w_{ij}(y) dy$$

$$= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} |J_{\phi_{\mathbf{e}}}| \int_{\Omega} K(\phi_{\mathbf{e}'}(t_{1m}), \phi_{\mathbf{e}}(y)) w_{1l}(y) dy.$$
(4.17)

For cases of $i \leq 1$, or $i' \leq 1$ or both, computation of $K_{i'j',ij}$ is easier, and one can slightly modify (4.17) to obtain the desired formulae. Note, if K(x, y) is a weakly singular kernel with discontinuity at x = y, in view of the integral in (4.17) the singular points of its integrand are at

$$y_s = \phi_{\mathbf{e}}^{-1} \phi_{\mathbf{e}'}(t_{1m}), \quad m \in \mathbb{Z}_{12}.$$
 (4.18)

We point out that the singular point y is not necessary in the domain Ω . Following lemma addresses this issue.

Lemma 4.5 For those elements $K_{i'j',ij}$ on the main diagonal line (as defined in Definition 4.2) of the submatrix $\mathbf{K}_{i'i}$, with $i, i' \geq 2$, let K(x, y) be a weakly singular kernel with singularity at y = x,

- 1. when i' = i, the singular points are at $y_s = t_{1m}$, where $m \in \mathbb{Z}_{12}$;
- 2. when i > i' the singular points are at $y_s = \phi_{\mathbf{e}_2}^{-1}(t_{1m})$, where $\mathbf{e}_2 \in \mathbb{Z}_{i-i'}$. Note, in this case y is not necessarily in Ω ;
- 3. when i < i' the singular points are at $y_s = \phi_{\mathbf{e}'_2}(t_{1m})$, where $\mathbf{e}'_2 \in \mathbb{Z}_{i'-i}$. In this case y is always in Ω .

For all other elements off the main diagonal line in each block $\mathbf{K}_{i'i}$, the integrals in (4.17) has no singular points.

Proof: We first prove the statements regarding elements on the main diagonal lines in each submatrix $\mathbf{K}_{i'i}$ with $i', i \geq 2$. Again we assume $j = \mu(\mathbf{e})r + l$, and $j' = \mu(\mathbf{e}')r + l'$, where, $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$, $\mathbf{e}' \in \mathbb{Z}_{\mu}^{i'-1}$, and $l, l' \in \mathbb{Z}_r$.

When
$$i' = i$$
, $\mathbf{e} = \mathbf{e}'$, so $y_s = t_{1m}$ by (4.18)

When i > i', let $\mathbf{e} = (\mathbf{e}_1, \mathbf{e}_2)$, where $\mathbf{e}_1 = (e_0, \dots, e_{i'-2})$ and $\mathbf{e}_2 = (e_{i'-1}, \dots, e_{i-2})$. For the elements on the main diagonal of $\mathbf{K}_{i'i}$, $\mathbf{e}' = \mathbf{e}_1$. Hence,

$$y_s = (\phi_{\mathbf{e}_1}\phi_{\mathbf{e}_2})^{-1}\phi_{\mathbf{e}'}(t_{1m}) = \phi_{\mathbf{e}_2}^{-1}\phi_{\mathbf{e}_1}^{-1}\phi_{\mathbf{e}'}(t_{1m}) = \phi_{\mathbf{e}_2}^{-1}(t_{1m}).$$
(4.19)

As i > i', the length of $\mathbf{e}_2 \ge 1$. Since $\phi_{\mathbf{e}_2}$ is a contraction, so in this case y_s may be outside of Ω .

When
$$i < i'$$
, let $\mathbf{e}' = (\mathbf{e}'_1, \mathbf{e}'_2)$, where, $\mathbf{e}'_1 = (e'_0, \dots, e'_{i-2})$, and $\mathbf{e}'_2 = (e'_{i-1}, \dots, e'_{i'-2})$. Then,
 $y_s = \phi_{\mathbf{e}}^{-1} \phi_{\mathbf{e}'_1} \phi_{\mathbf{e}'_2}(t_{1m}) = \phi_{\mathbf{e}'_2}(t_{1m})$.

since
$$\mathbf{e} = \mathbf{e}'_1$$
. Note in this case y_s is always inside Ω since $\phi_{\mathbf{e}'_2}$ is a contractive mapping

For elements which are off the main diagonal lines of $\mathbf{K}_{i'i}$, $\hat{S}_{i'j'}$ and S_{ij} which correspond to the supports of the pair $(\ell_{i'j'}, w_{ij})$ are disjoint except possible common boundary points. Recall that all the collocation points are interior points of $\hat{S}_{i'j'}$, therefore there are no singular points inside the domain of integration for integrals associated with those elements. Although we can compute elements of \mathbf{K}_n by (4.17) and Theorem 4.4, when singular points are outside Ω , it's not easy for us to pick up appropriate meshes to compute quadratures. Especially when singular points are very close to Ω but remaining outside of Ω . In this case if we choose equally spaced meshes to subdivide domains of integration, it is obvious that it would be improper. If we intend to choose unequally spaced meshes, the question of how to uniformly deal with all four cases stated in Lemma 4.5 occurs. Next Theorem provides ways to uniformly treat all different types of integrals associated with elements of \mathbf{K}_n with uniform accuracy.

Theorem 4.6 Let $P_k, k \in \mathbb{Z}_3$ be vertices of the original triangular domain Ω supported $w_{1l}, l \in \mathbb{Z}_r$, where P_0 is the vertex of the right angle, and P_1, P_2 are indexed counterclockwisely (see Figure 4.4). Denote new vertices of the transformed domain $\phi_{\mathbf{e}}(\Omega)$ by \hat{P}_k , and $\hat{P}_k = \phi_{\mathbf{e}}(P_k), k \in \mathbb{Z}_3$, where $\mathbf{e} \in \mathbb{Z}_{\mu}^{i-1}$ for $i \geq 2$. Denote coordinates of \hat{P}_k by (\hat{x}_k, \hat{y}_k) . For $i, i' \geq 2$, the elements $K_{i'j',ij}$ of \mathbf{K}_n can be computed by the following way:

$$K_{i'j',ij} = \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\hat{x}_0}^{\hat{x}_1} \int_{\hat{y}_0}^{-x + (\hat{x}_1 + \hat{y}_1)} K(\phi_{\mathbf{e}'}(t_{1m}), \eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}y \mathrm{d}x$$
(4.20)

where, t_{1m} , η , P_k , $\hat{P}_k \in \mathbb{R}^2$, $k \in \mathbb{Z}_3$, and $\eta = (x, y)$.

Proof: Since $\phi_{\mathbf{e}}$ is a contractive affine mapping and also conformal, $\phi_{\mathbf{e}}(\Omega)$ is a right triangle too. The orientation of $\phi_{\mathbf{e}}(\Omega)$ is either as the lower left one in Fig. 4.4 or as the lower right one in Fig. 4.4.

Case 1: The new domain $\phi_{\mathbf{e}}(\Omega)$ is as the lower left one in Fig. 4.4. Note the hypotenuse $\hat{P}_1\hat{P}_2$ always has slope -1 and it's equation is

$$y = -x + (\hat{x}_1 + \hat{y}_1).$$

Hence

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$$\begin{aligned}
K_{i'j',ij} &= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\phi_{\mathbf{e}}(\Omega)} K(\phi_{\mathbf{e}'}(t_{1m}), \eta) w_{ij}(\eta) \mathrm{d}\eta \\
&= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\phi_{\mathbf{e}}(\Omega)} K(\phi_{\mathbf{e}'}(t_{1m}), \eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}\eta \\
&= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\hat{x}_0}^{\hat{x}_1} \int_{\hat{y}_0}^{-x + (\hat{x}_1 + \hat{y}_1)} K(\phi_{\mathbf{e}'}(t_{1m}), \eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}y \mathrm{d}x
\end{aligned} \tag{4.21}$$

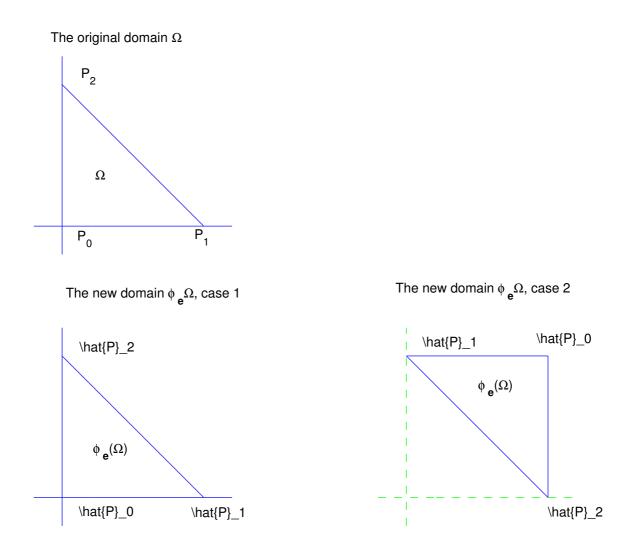


Figure 4.4: Domains of integration

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Case 2: The new domain $\phi_{\mathbf{e}}(\Omega)$ is as the lower right one in Fig. 4.4.

$$\begin{split} K_{i'j',ij} &= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\phi_{\mathbf{e}}(\Omega)} K(\phi_{\mathbf{e}'}(t_{1m}),\eta) w_{ij}(\eta) \mathrm{d}\eta \\ &= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\phi_{\mathbf{e}}(\Omega)} K(\phi_{\mathbf{e}'}(t_{1m}),\eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}\eta \\ &= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\hat{x}_1}^{\hat{x}_0} \int_{-x+(\hat{x}_1+\hat{y}_1)}^{\hat{y}_0} K(\phi_{\mathbf{e}'}(t_{1m}),\eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}y \mathrm{d}x \\ &= \sum_{m \in \mathbb{Z}_{12}} d_{l'm} \int_{\hat{x}_0}^{\hat{x}_1} \int_{\hat{y}_0}^{-x+(\hat{x}_1+\hat{y}_1)} K(\phi_{\mathbf{e}'}(t_{1m}),\eta) w_{1l}(\phi_{\mathbf{e}}^{-1}(\eta)) \mathrm{d}y \mathrm{d}x. \end{split}$$

bletes the proof.

This completes the proof.

Remark: when $i \leq 1$ or $i' \leq 1$ or both, it is easier to compute $K_{i'j',ij}$. Slightly modify (4.20) if necessary.

Since the point $\phi_{\mathbf{e}'}(t_{1m}), m \in \mathbb{Z}_{12}$ is always contained in Ω , (4.20) allows an easy way to get an unequally spaced mesh on Ω with $\phi_{\mathbf{e}'}(t_{1m})$ as the singular point. This will be detailly discussed in Section 4.3. Theorem 4.6 will be the actual method we adopt to compute elements of \mathbf{K}_n . Fig. 4.5 depicts the computed full matrix \mathbf{K}_3 from example in solving (6.1). For viewing effect, absolute values of matrix entries are drawn. After the wavelet transform, the matrix \mathbf{K}_n is amazingly numerically sparse. Namely many entries are relatively very small, and the matrix appears as "finger" shape.

4.3An Adaptive Quadrature Rule

All the elements of \mathbf{K}_n need to compute

$$\int_{a}^{b} \int_{c}^{d-y_{1}} K(x,y) w_{ij}(y) dy_{2} dy_{1}, \quad 0 \le a, b, c, d \le 1,$$
(4.22)

where, $x, y \in \mathbb{R}^2$ with $x = (x_1, x_2), y = (y_1, y_2)$. $w_{ij}, i \in \mathbb{N}_0, j \in \mathbb{Z}_{w(i)}$ is a wavelet function. K(x,y) is a weakly singular function on $\Omega \times \Omega$. Namely, for $x, y \in \Omega, x \neq y$, the kernel K has continuous partial derivatives $D_x^{\alpha} D_y^{\beta} K(x, y)$ for $|\alpha| \leq k$, $|\beta| \leq k$. Moreover, there exists positive constants σ and θ_1 with $\sigma < 2$ such that for $|\alpha| = |\beta| = k$ there holds

$$\left|D_x^{\alpha} D_y^{\beta} K(x,y)\right| \le \frac{\theta_1}{|x-y|^{\sigma+|\alpha|+|\beta|}}.$$
(4.23)

While K(x, y) is singular when x = y. Note w_{ij} for $i \ge 1, j \in \mathbb{Z}_{w(i)}$ is a piecewise polynomial and has shrinking supports as i increases. In general, there are two different cases:

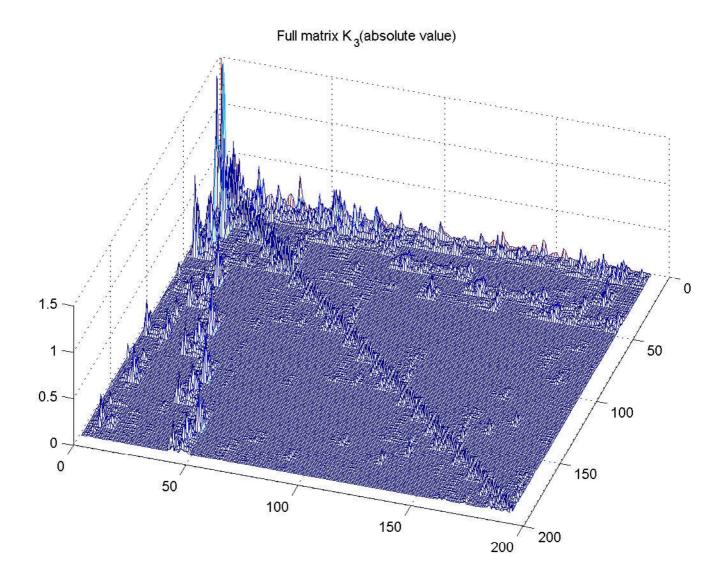


Figure 4.5: Full Matrix \mathbf{K}_3

- 1. the singular point x is within the support of w_{ij} ,
- 2. x is not in the support of w_{ij} .

We want to have a uniform scheme to deal with these two cases and the shrinking supports of w_{ij} , nevertheless, uniform accuracies are desired for all cases. [KX] provides a good quadrature method for 1-D integrals with weakly singular kernels. we adapt the idea proposed therein to develop an efficient adaptive algorithm to compute double integrals equipped with weakly singular kernels. Rewrite (4.22) into

$$I := \int_{a}^{b} h(x; y_1) \mathrm{d}y_1, \tag{4.24}$$

where, $h(x; y_1) := \int_c^{d-y_1} K(x, y) w_{ij}(y) dy_2$. Denote $\max\{x_1, 1 - x_1\}$ by c_{max} and $\min\{x_1, 1 - x_1\}$ by c_{min} . Note $1 \ge c_{max} \ge \frac{1}{2}$ and $0 \le c_{min} \le \frac{1}{2}$. We first obtain an uneven partition of [a, b] by the following **Algorithm**:

1. Subdivide $[0, c_{max}]$ into n subintervals by the following way:

$$\hat{t}_j = c_{max} \left(\frac{j}{n}\right)^q, \quad j \in \mathbb{Z}_{n+1},$$
(4.25)

where n is determined by

$$n = \lfloor m \frac{c_{max}}{1/2} = 2mc_{max} \rfloor.$$
(4.26)

Note $\hat{t}_0 = 0$ and $\hat{t}_n = c_{max}$. *m* is a prescribed integer. $q = \frac{2k+1}{1-\alpha}$ (see [KX] for detail). *k* is the total degree of the interpolatory piecewise polynomial used in the compound quadrature. For a function $g, \alpha \in (0, 1)$ is called the index of singularity. It's defined by

$$|g^{(\beta)}(x)| \le C|x|^{(-\alpha-\beta)}, \text{ for } x \ne 0.$$

This notation was first introduced by Rice in [Ri].

2. Determine the number of subintervals J in the shorter interval $[0, c_{min}]$. Require that

$$c_{max} \cdot \left(\frac{J}{n}\right)^q = c_{min}.\tag{4.27}$$

By (4.27), we can take

$$J = \left\lceil \left(\frac{c_{min}}{c_{max}}\right)^{\frac{1}{q}} n \right\rceil.$$
(4.28)

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3. if $x_1 \le 1 - x_1$ Set $t'_0 = 0$. Let

$$t'_{J-j} = x_1 - \hat{t}_j, \quad j \in \mathbb{Z}_J,$$
 (4.29)

and

$$t'_{J+i} = x_1 + \hat{t}_i, \quad i \in \mathbb{Z}_{n+1}.$$
 (4.30)

Note $t'_{J} = x_1$, and $t'_{J+n} = 1$.

4. if $x_1 > 1 - x_1$

$$t'_{n-j} = x_1 - \hat{t}_j, \quad j \in \mathbb{Z}_{n+1},$$
(4.31)

and

$$t'_{n+i} = x_1 + \hat{t}_i, \quad i \in \mathbb{Z}_J.$$
 (4.32)

Note $t'_0 = 0, t'_n = x_1$ and we set $t'_{n+J} = 1$.

Define that $\triangle_{t'}^{y_1} := \{t'_i : i \in \mathbb{Z}_{n+J+1}\}$. Let $\triangle(w_{ij})$ be the set of discontinuous points of w_{ij} along y_1 -direction. Define

$$\Delta_a^b := \Delta_{t'}^{y_1} \cup \Delta(w_{ij}) \cup \{a, b\} \cap [a, b].$$

Reorder the points in $riangle_a^b$ by an increasing order and rename them as

$$a = t_0 < t_1 < \dots < t_{m_1} = b_1$$

This is an desired partition of [a, b]. Now (4.24) becomes into

$$I = \sum_{i=0}^{m_1-1} \int_{t_i}^{t_{i+1}} h(x;\xi) d\xi = \int_{-1}^1 H(s) ds, \qquad (4.33)$$

where,

$$H(s) = \sum_{i=0}^{m_1-1} \bar{t}_i^- h\left(x; \bar{t}_i^- s + \bar{t}_i^+\right), \qquad (4.34)$$

with

$$\bar{t}_i^+ = \frac{t_{i+1} + t_i}{2}, \quad \bar{t}_i^- = \frac{t_{i+1} - t_i}{2}.$$
(4.35)

Using Gaussian quadrature method with the interpolating polynomial of degree k', we obtain the quadrature value of I

$$I_m = \sum_{j \in \mathbb{Z}_{k'}} w_j H(u_j), \tag{4.36}$$

where, $\{u_i\}, i \in \mathbb{Z}_{k'}$ are the k' zeros of the Legendre polynomial of degree k' on (-1, 1), and

$$w_i = \int_{-1}^{1} l_i(x) dx \quad \text{with} \quad l_i(x) = \prod_{j \neq i} \frac{x - u_j}{u_i - u_j}, \quad i \in \mathbb{Z}_{k'}$$

Remark: We set $h(x;\xi) = 0$ if $\xi \in [t'_{J-1}, t'_{J+1})$ when $x_1 \le 1 - x_1$ or if $\xi \in [t'_{n-1}, t'_{n+1})$ when $x_1 > 1 - x_1$.

We need to evaluate $H(u_e), e \in \mathbb{Z}_{k'}$. To this end, in turn we shall need to compute $h(x; y_{ie})$, where

$$y_{ie} = \bar{t}_i^- u_e + \bar{t}_i^+$$

Now

$$h(x; y_{ie}) = \int_{c}^{d-y_{ie}} K(x; (y_{ie}, \eta)) f(y_{ie}, \eta) d\eta$$

:= $\int_{c}^{d'} g(x; (y_{ie}, \eta)) d\eta,$ (4.37)

where

$$g(x;(y_{ie},\eta)) = K(x;(y_{ie},\eta))f(y_{ie},\eta).$$
(4.38)

We use exactly the same idea as to compute I to compute $h(x; y_{ie})$. Denote the quadrature value of $h(x; y_{ie})$ by $h_m(x; y_{ie})$. Let $\triangle_c^{d'} := \{c = s_1 < s_2 < \cdots < s_{m_2}\}$ obtained similarly as we obtain \triangle_a^b , with x_2 as the shifting parameter. Then

$$h(x; y_{ie}) := \sum_{j=0}^{m_2-1} \int_{s_j}^{s_{j+1}} g(x; (y_{ie}, \xi)) d\xi = \int_{-1}^1 G(\xi) d\xi, \qquad (4.39)$$

with,

$$G(\xi) = \sum_{j=0}^{m_2-1} \bar{s}_j^- g\left(x; (y_{ie}, \bar{s}_j^- \xi + \bar{s}_j^+)\right)$$
(4.40)

Again, using Gaussian quadrature method with the interpolating polynomial of degree k', we obtain

$$h_m(x; y_{ie}) = \sum_{j \in \mathbb{Z}_{k'}} w_j G(u_j),$$
 (4.41)

where, $G(u_e), e \in \mathbb{Z}_{k'}$ are computed by (4.40).

Remark: We set $g(x; (y_{ie}, \xi)) = 0$ if $\xi \in [s'_{J-1}, s'_{J+1})$ when $x_2 \le 1 - x_2$ or if $\xi \in [s'_{n-1}, s'_{n+1})$

when $x_2 > 1 - x_2$, where, $\triangle_{s'}^{y_2} := \{s'_j : j \in \mathbb{Z}_{n+J+1}\}$ is similarly obtained as $\triangle_{t'}^{y_1}$.

The numerical quadrature rule proposed here is extremely efficient, in the sense that not only it can automatically adjust the degree of fineness for the mesh regarding different domains of integration, but also use less and less grid points to compute the quadrature value when the domain is getting smaller and smaller. Nevertheless, in all cases it gives exactly same accuracies. We next show that the numerical integral I_m computed by (4.36) for the weakly singular double integral I in (4.22) is of order 2k, i.e.,

$$|I_m - I| = \mathcal{O}(\delta_m^{2k}), \qquad (4.42)$$

where

$$\delta_m := \min\left(\frac{1}{2m}, \max(b-a, d-c)\right).$$

4.4 Error Analysis of the Adaptive Quadrature Rule

Let us first recall for $F \in C^{2k}(0,1)$,

• The *k*-point Gauss-Legendre rule:

$$\int_{-1}^{1} F(x)dx = \sum_{i \in \mathbb{Z}_{k}} w_{i}F(u_{i}) + C_{k}F^{(2k)}(\epsilon_{c}), \qquad (4.43)$$

where ϵ_c is some number in (-1, 1), C_k is a constant independent of F, $\{u_i\}, i \in \mathbb{Z}_k$ are k zeros of the Legendre polynomial of degree k on (-1, 1), and

$$w_i = \int_{-1}^{1} l_i(x) dx \quad \text{with} \quad l_i(x) = \prod_{j \neq i} \frac{x - u_j}{u_i - u_j}, \quad i \in \mathbb{Z}_k$$

Next we establish a composite k-point Gauss-Legendre rule for singular integrals with m non-uniform subdivisions (see [KX] for more related results).

Lemma 4.7 Suppose that f defined on (0,1) with

$$|f^{(2k)}(x)| \le Cx^{-\alpha-2k}, \text{ for all } x \in (0,1) \text{ and some } \alpha \in (0,1).$$
 (4.44)

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Then, for any given integer m > 0, there holds

$$\int_{0}^{1} f(x)dx = \sum_{i \in \mathbb{Z}_{k}} w_{i} \left(\sum_{j=1}^{m-1} \bar{x}_{j}^{-} f(\bar{x}_{j}^{-} u_{i} + \bar{x}_{j}^{+}) \right) + O\left(\frac{1}{m^{2k}}\right), \tag{4.45}$$

where

$$x_j = \left(\frac{j}{m}\right)^q, \quad q = \frac{2k+1}{1-\alpha},$$

for $j \in \mathbb{Z}_{m+1}$.

Proof: It follows from (4.44) that $|f(x)| \leq Cx^{-\alpha}$ for $x \in (0, 1)$, and thus we have

$$\left| \int_{0}^{x_{1}} f(x) dx \right| \le C x_{1}^{1-\alpha} = C \frac{1}{m^{2k+1}}.$$
(4.46)

Since

$$\int_{0}^{1} f(x)dx = \int_{0}^{x_{1}} f(x)dx + \int_{x_{1}}^{1} f(x)dx, \qquad (4.47)$$

and

$$\int_{x_1}^1 f(x) = \sum_{j=1}^{m-1} \int_{x_j}^{x_{j+1}} f(x) dx = \int_{-1}^1 F(x) dx, \qquad (4.48)$$

where

$$F(x) = \sum_{j=1}^{m-1} \bar{x}_j^- f\left(\bar{x}_j^- x + \bar{x}_j^+\right), \qquad (4.49)$$

Hence, applying the k-point rule (4.43) to (4.48), yields

$$\int_{x_1}^1 f(x)dx = \int_{-1}^1 F(x)dx = \sum_{i \in \mathbb{Z}_k} w_i F(u_i) + C_k F^{(2k)}(\epsilon_c).$$
(4.50)

Note the first term on the right side of (4.50) is exactly the first term on the right side of (4.45), therefore it suffices to verify that the second term on the right side of (4.50) is of $O\left(\frac{1}{m^{2k}}\right)$. From (4.49),

$$F^{(2k)}(x) = \sum_{j=1}^{m-1} \left(\bar{x}_j^-\right)^{2k+1} f^{(2k)} \left(\bar{x}_j^- x + \bar{x}_j^+\right),$$

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and thus, by condition (4.44) on $f^{(2k)}$,

$$|F^{(2k)}(\epsilon_{c})| \leq \sum_{j=1}^{m-1} (\bar{x}_{j}^{-})^{2k+1} |f^{(2k)}(\bar{x}_{j}^{-}\epsilon_{c} + \bar{x}_{j}^{+})|$$

$$\leq C \sum_{j=1}^{m-1} (2\bar{x}_{j}^{-})^{2k+1} (x_{j})^{-\alpha-2k}$$

$$= \frac{C}{m^{q(1-\alpha)}} \sum_{j=1}^{m-1} ((j+1)^{q} - (j)^{q})^{2k+1} (j)^{-q(\alpha+2k)}$$

$$\leq \frac{C}{m^{q(1-\alpha)}} \sum_{j=1}^{m-1} q^{2k+1} 2^{(q-1)(2k+1)} (j)^{(q-1)(2k+1)} (j)^{-q(\alpha+2k)}$$

$$= \frac{C}{m^{2k+1}} \sum_{j=1}^{m-1} 1 \leq \frac{C}{m^{2k}},$$
(4.51)

where we have used the inequality

$$(j+1)^q - (j)^q = q\xi^{q-1} \le q(j+1)^{q-1} \le q2^{q-1}(j)^{q-1}$$
(4.52)

with some $\xi \in (j, j + 1)$ and $q \ge 1$. Combining (4.47), (4.46), (4.50) and (4.51), we obtain (4.45).

Theorem 4.8 Suppose that f defined on $(0, 1) \times (0, 1)$ satisfies

$$\left|\frac{\partial^{2k}f}{\partial x^{2k}}(x,y)\right| \le Cx^{-\alpha-2k}, \quad \left|\frac{\partial^{2k}f}{\partial y^{2k}}(x,y)\right| \le Cy^{-\alpha-2k} \tag{4.53}$$

for all $(x,y) \in (0,1) \times (0,1)$ and some $\alpha \in (0,1)$. Let $\beta \in C^{2k}[0,1]$ satisfying

$$0 \le \beta(x) \le 1$$
 on $(0,1)$. (4.54)

Then, for any given integer m > 0, there holds

$$\int_{0}^{1} \int_{0}^{\beta(x)} f(x,y) dy \, dx = \sum_{i \in \mathbb{Z}_{k}} w_{i} \left(\sum_{j=1}^{m-1} \bar{x}_{j}^{-} F_{ij} \right) + O\left(\frac{1}{m^{2k}} \right),$$

$$with \ F_{ij} = \sum_{\mu \in \mathbb{Z}_{k}} w_{\mu} \left(\sum_{\nu=1}^{m-1} \bar{y}_{\nu}^{-} f(v_{ij}, \bar{y}_{\nu}^{-} u_{\mu} + \bar{y}_{\nu}^{+}) \right),$$
(4.55)

where

$$x_j = \left(\frac{j}{m}\right)^q, \quad j \in \mathbb{Z}_{m+1}, \quad q = \frac{2k+1}{1-\alpha},$$

$$v_{ij} = \bar{x}_j^- u_i + \bar{x}_j^+, \quad m_{ij} = \left[(\beta(v_{ij}))^{\frac{1}{q}} m \right],$$
$$y_{\nu} = \left(\frac{\nu}{m}\right)^q, \quad \nu \in \mathbb{Z}_{m_{ij}} \quad and \quad y_{m_{ij}} = \beta(v_{ij})$$

Moreover the computational complexity $\mathcal{N}(I_m)$ is

$$\mathcal{N}(I_m) \le Cm^2,\tag{4.56}$$

where, $\mathcal{N}(I_m)$ is the number of the point evaluations of the integrand f, and I_m is the first term in (4.55), namely the numerical approximation of the integral.

Proof: Let

$$F(x) = \int_0^{\beta(x)} f(x, y) \, dy$$

Note

 $F^{(2k)}(x) = \int_0^{\beta(x)} \frac{\partial^{2k} f}{\partial x^{2k}}(x, y) dy + \text{combination of lower partial derivatives of } f \text{ with respect to } x.$

By (4.53)-(4.54), we have that

$$|F^{(2k)}(x)| \le Cx^{-\alpha - 2k}$$
 for $x \in (0, 1)$.

From Lemma 4.7,

$$\int_{0}^{1} \int_{0}^{\beta(x)} f(x,y) \, dy dx = \int_{0}^{1} F(x) dx = \sum_{i \in \mathbb{Z}_{k}} w_{i} \left(\sum_{j=1}^{m-1} \bar{x}_{j}^{-} F(v_{ij}) \right) + O\left(\frac{1}{m^{2k}} \right). \tag{4.57}$$

Each evaluation of $F(v_{ij})$ requires another quadrature. Using Lemma 4.7 (4.45) again yields:

$$F(v_{ij}) = \int_0^{\beta(v_{ij})} f(v_{ij}, y) \, dy = F_{ij} + O\left(\frac{1}{m^{2k}}\right)$$
(4.58)

It remains to estimate the sum

$$E_{ij} = \sum_{j=1}^{m-1} \bar{x}_j^- \frac{1}{m^{2k}}.$$

While

$$\begin{aligned} |E_{ij}| &\leq 2\sum_{j=1}^{m-1} 2\bar{x}_j^- \frac{1}{m^{2k}} = \frac{2}{m^{2k}} \sum_{j=1}^{m-1} \frac{(j+1)^q - j^q}{m^q} \\ &\leq \frac{C}{m^{2k}} \sum_{j=1}^{m-1} \frac{j^q}{m^q} \leq \frac{C}{m^{2k}}, \end{aligned}$$
(4.59)

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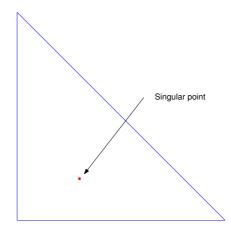


Figure 4.6: Domain of integration for Example 4.9

where we have used the inequality (4.52) again. Therefore, (4.55) follows from (4.57)-(4.59). In view of (4.55), it is clear that (4.94) holds.

We now present several numerical examples using above scheme. We shall consider $K(x, y) = \frac{1}{x-y}$. The Gauss quadrature with interpolating polynomial of order k = 2 is employed. Consequently, we choose q = 5 for the non-uniform partition. The theoretical order of accuracy (2k = 4) by this quadrature rule is verified in Example 4.9 and Example 4.10. The convergent rate is computed by $\log_2 \frac{|I_m - I_{2m}|}{|I_{2m} - I_{4m}|}$ in these two examples.

Example 4.9 Consider following integral

$$I = \int_0^1 \int_0^{1-x} \frac{x+y+1}{\sqrt{(x-0.3)^2 + (y-0.2)^2}} \mathrm{d}y \mathrm{d}x$$
(4.60)

In this example the singular point (0.3, 0.2) is inside the domain of integration-the unit triangle Ω (see Fig. 4.6). We tabulate our computation results in Table 4.1. It is clearly shown that the quadrature rule gives accurate results, and its order of convergence is also verified.

Example 4.10 Consider following integral

$$I = \int_{0.5}^{1} \int_{0}^{1-x} \frac{x+y+1}{\sqrt{(x-0.3)^2 + (y-0.2)^2}} \mathrm{d}y \mathrm{d}x.$$
 (4.61)

	r	1	
m	Numerical I_m	$ I_m - I_{2m} $	Convergent rate
5	3.714414251744		
10	3.721811162863	7.3969e-03	
20	3.722373749843	5.6259e-04	3.7168
40	3.722412249476	3.8500e-05	3.8692
80	3.722414764040	2.5146e-06	3.9365
160	3.722414924162	1.6012 e-07	3.9731
320	3.722414934272	1.0110e-08	3.9853
640	3.722414934906	6.3451e-10	3.9940
1280	3.722414934946	3.9740e-11	3.9970
2560	3.722414934948	2.4864e-12	3.9985

Table 4.1: Quadrature computing results for Example 4.9

Note, in this example, the singular point (0.3, 0.2) is outside the domain of integration (see Fig. 4.7). We tabulate the computing results in Table 4.2.

Example 4.11 In this example, we test if our quadrature rule is good enough in solving the actual integral equation (6.1), the linear system of equations are solved by standard Gaussian elimination method. No compression is assumed. The results are tabulated in Table 4.3. Order of convergence is computed by $\log_2 \frac{\|u-u_n\|_{\infty}}{\|u-u_{n+1}\|_{\infty}}$, where n stands for level number of the resolution, u is the exact solution and u_n is the approximation to u in space \mathbb{F}_n ; The theoretical convergent rate is 2. We can see from the results that results by using our quadrature rule are very close to those obtained without quadrature rule.

Remark: The adaptive strategy of the proposed quadrature rule can be applied to any similar case.

Following theorem tells us that the adaptive strategy dramatically reduces the computational complexity for integration.

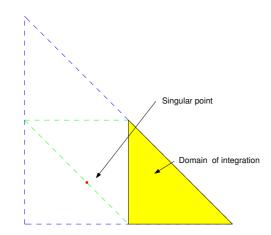


Figure 4.7: Domain of integration for Example 4.10

m	Numerical I_m	$ I_m - I_{2m} $	Convergent rate
5	0.6337498001541		
10	0.6344202251264	6.7042e-04	
20	0.6344713062075	5.1081e-05	3.7142
40	0.6344745338648	3.2277e-06	3.9842
80	0.6344747360404	2.0218e-07	3.9968
160	0.6344747488410	1.2801e-08	3.9813
320	0.6344747496463	8.0531e-10	3.9905
640	0.6344747496970	5.0714e-11	3.9891
1280	0.6344747497002	3.1701e-12	3.9998
2560	0.6344747497004	1.9855e-13	3.9970

Table 4.2: Quadrature computing results for example 4.10

	without qua	drature	with quadrature			
n	L^{∞} -error	order	m	L^{∞} -error	order	
1	1.513187e-1		40	1.513184e-1		
2	3.599605e-2	2.0717	40	3.599379e-2	2.0718	
3	7.499082e-3	2.2631	40	7.496564e-3	2.2634	
4	1.849162e-3	2.0198	40	1.850269e-3	2.0185	
5	4.650610e-4	1.9914	40	4.657219e-4	1.9902	

Table 4.3: Comparison of uncompressed solutions by the wavelet collocation method, with and without numerical quadratures for matrix entries.

Theorem 4.12 In solving (2.2), if the basis functions of the approximation space are chosen as constructed in Chapter 3, and if the adaptive quadrature rule proposed in this section is employed, the complexity to compute all matrix entries in the discrete linear system is:

$$\mathcal{N}(I_m) \le Cf(n)\log f(n)m^2,\tag{4.62}$$

where, $\mathcal{N}(I_m)$ is the number of function evaluations of integrands, f(n) is the dimension of the approximation space, and n is a resolution level of the approximation space.

Proof: The proof is done by direct counting. For any row $i'j', (i', j') \in \mathbb{U}_n$, in view of (2.19), each entry in this row is computed through a w_{ij} , for $(i, j) \in \mathbb{U}_n$. We divide the elements in row i'j' into (n + 1) subgroups according to their i value. Namely, the elements of row i'j' is divided into subgroups

$$K_{i'j',0j}, \quad j \in \mathbb{Z}_{w(0)},$$
$$K_{i'j',1j}, \quad j \in \mathbb{Z}_{w(1)},$$
$$\vdots \qquad \vdots$$
$$K_{i'j',nj}, \quad j \in \mathbb{Z}_{w(n)}.$$

When $i = 0, w_{0j}, j \in \mathbb{Z}_{w(0)}$ are all supported on Ω . There are w(0) elements to be evaluated through integration, and each integration needs Cm^2 function evaluations by Theorem 4.8. Hence the number of function evaluations is

$$\mathcal{N}_0(I_m) = Cw(0)m^2.$$

When $i = 1, w_{1j}, j \in \mathbb{Z}_{w(1)}$ are all supported on Ω . The number of function evaluations is (note w(1) = r)

$$\mathcal{N}_1(I_m) = C.w(1)m^2 = C.rm^2.$$

For $1 < i \leq n$, since the supports of basis functions of \mathbb{W}_n satisfies (2.7) and (2.8), then the number of function evaluations of the integrand is

$$\mathcal{N}_i(I_m) = C.rm^2.$$

Thus,

$$\mathcal{N}(I_m) = f(n)(\mathcal{N}_0(I_m) + \mathcal{N}_1(I_m) + \dots + \mathcal{N}_n(I_m))$$

= $f(n)(Cw(0)m^2 + Crm^2 + Crm^2 + \dots + Crm^2)$
= $f(n)(Cw(0)m^2 + n \cdot Crm^2)$
 $\leq Cf(n)\log f(n)m^2.$

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4.5 Compression Schemes-Distance Computing

The matrix $\mathbf{K}_{\mathbf{n}}$ in general is non-symmetric and numerical sparse. [CMX3] gives following theoretical truncation criterion.

Let $\tilde{\mathbf{K}}_n := [\tilde{K}_{i'j',ij} : (i',j'), (i,j) \in \mathbb{U}_n]$ be the truncation matrix whose entries are defined in terms of a matrix truncation parameter $\epsilon_{i'i}$ by

$$\tilde{K}_{i'j',ij} := \begin{cases} K_{i'j',ij} & \text{if } \operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) \leq \epsilon_{i'i}, \\ 0 & \text{otherwise,} \end{cases}$$
(4.63)

where dist $(S_{i'j'}, S_{i,j})$ denotes the geometric distance between the support $\hat{S}_{i'j'}$ of a collocation functional $\ell_{i'j'}$ and the support S_{ij} of a wavelet function w_{ij} . The truncation parameter $\epsilon_{i'i}$ is chosen such that for some positive constants a and $\nu > 1$,

$$\epsilon_{i'i} = \max\{a\mu^{[-i+b'(n-i')]/2}, \ \nu(d_i + d_{i'})\} \quad i, i' \in \mathbb{Z}_{n+1},$$
(4.64)

where,

$$\frac{k - \sigma'}{2k - \sigma'} < b' \le 1, \quad \text{with } 0 < \sigma' < 2 - \sigma.$$

strategies		Level n						
	2	3	4	5	6	7	8	9
strategy $1(\nu = 1)$	1.0000	0.9780	0.6766	0.3172	0.1198	0.0408	0.0135	0.0045
strategy $2(\nu = 1)$	1.0000	0.8945	0.5202	0.2276	0.0857	0.0302	0.0105	0.0037
strategy $3(\nu = 1)$	1.0000	0.8857	0.4488	0.1699	0.0572	0.0189	0.0066	0.0025
block truncation	1.0000	0.7583	0.4166	0.1810	0.0677	0.0230	0.0073	0.0022
$(p_{i'i} = \sqrt{3})$								

Table 4.4: Comparison of compression rates of different strategies

 $d_i := O(\mu^{-(i/2)})$ is the diameter of S_{ij} , $j \in \mathbb{Z}_{w(i)}$. n is the level number of the approximation space. Note the truncating parameter $\epsilon_{i'i}$ (4.64) changes as i, i' change. We present here several different *practical* strategies. Three strategies are investigated in this section. For strategy 1 we choose

$$\epsilon_{i'i} = \nu(d_{i'} + d_i). \tag{4.65}$$

For strategy 2 we choose

$$\epsilon_{i'i} = \nu \max\{d_{i'}, d_i\}. \tag{4.66}$$

For strategy 3 we choose

$$\epsilon_{i'i} = \nu \min\{d_{i'}, d_i\}.$$
 (4.67)

A comparison of compression rates of different strategies is depicted in Fig. 4.8, their actual numbers are also listed in Table 4.4 (We take $\nu = 1$ for strategy 1, 2, 3). The compression rate is computed as the ratio of the number of nonzero entries in the compressed matrix $\tilde{\mathbf{K}}_n$ over the total number of entries of \mathbf{K}_n . Observe that the compression rate of strategy 2 is in between of those of strategy 1 and strategy 3.

Moreover, we test these different strategies solving the integral equation (6.1) with $\nu =$ 1. In order to exclude other affecting factors, we use standard Gaussian elimination to solve the resulting system. All the matrix entries are computed exactly. The results are tabulated in Table 4.5, Table 4.6 and Table 4.7. Order of convergence is computed by $\log_2 \frac{\|u-u_n\|_{\infty}}{\|u-u_{n+1}\|_{\infty}}$, where *n* stands for level number of the resolution, *u* is the exact solution and u_n is an approximation to *u* in space \mathbb{F}_n ; The theoretical convergent rate is 2. The

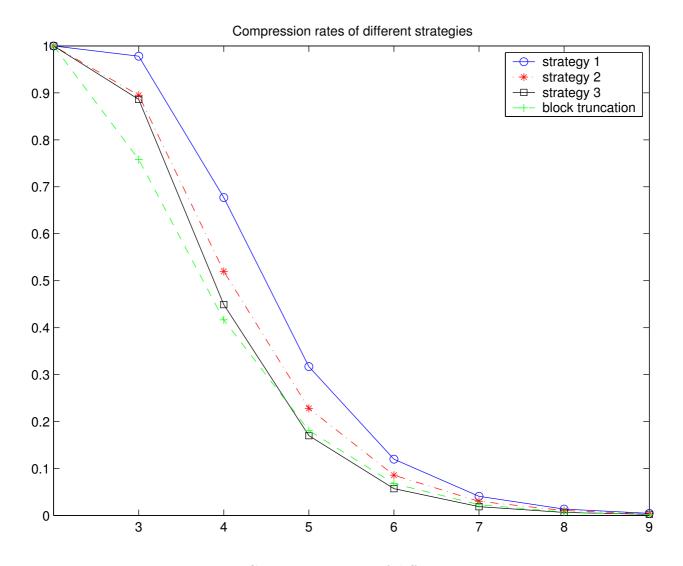


Figure 4.8: Compression rates of different strategies

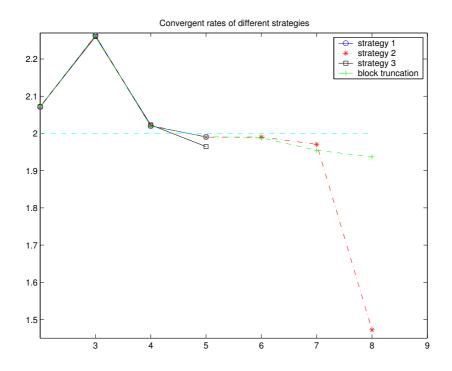


Figure 4.9: Convergent rates of different strategies

comparison of their convergent rates for different strategies is plotted in Figure 4.9. By comparing with the theoretical convergent rate 2, from Fig. 4.9 we can tell that strategy 1 is relatively conservative and strategy 3 is somewhat insufficient. While strategy 2 gives relatively satisfied results.

We further investigate computing results by choosing different parameter ν . The results are tabulated in Table 4.8, Table 4.9 and Table 4.10. The convergent rates of strategy 3 with different ν are depicted in Figure 4.10, and the convergent rates of strategy 2 with different ν are depicted in Figure 4.11. From those tabulated results and the two figures Fig. 4.10 and Fig. 4.11, we see that for the adopted strategy 2, $\nu = 1$ gives satisfied results. While smaller ν doesn't give satisfied results.

The matrix \mathbf{K}_n after compression is quite sparse when n is relatively large (See Fig. 4.12, Fig. 4.13 and Fig. 4.14. These sparse matrices are computed for integral equation (6.1). Special sparse matrix storage is used when doing the actual computing. Not only this speeds up computing, but also this dramatically saves computer memory. In fact when n

	full collocation	on solution u_n	compressed solution \tilde{u}_n			
n	L^{∞} -error	order	L^{∞} -error	order	compression rate	
1	1.513187e-1		1.513187e-1		1.000000	
2	3.599605e-2	2.0717	3.599605e-2	2.0717	1.000000	
3	7.499082e-3	2.2631	7.499309e-3	2.2630	0.978027	
4	1.849162e-3	2.0198	1.849060e-3	2.0200	0.676590	
5	4.650610e-4	1.9914	4.652252e-4	1.9908	0.317277	

Strategy 1: $\epsilon_{i'i} = \nu(d_{i'} + d_i)$

Table 4.5: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 1 with $\nu = 1$).

Strategy 2: $\epsilon_{i'i} = \nu \max\{d_{i'}, d_i\}$

	full collocation	on solution u_n	compressed solution \tilde{u}_n			
n	L^{∞} -error	order	L^{∞} -error	order	compression rate	
1	1.513187e-1		1.513187e-1		1.000000	
2	3.599605e-2	2.0717	3.599605e-2	2.0717	1.000000	
3	7.499082e-3	2.2631	7.501118e-3	2.2627	0.894531	
4	1.849162e-3	2.0198	1.847625e-3	2.0214	0.520172	
5	4.650610e-4	1.9914	4.650882e-4	1.9901	0.227601	

Table 4.6: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 2 with $\nu = 1$).

	full collocation	on solution u_n	compressed solution \tilde{u}_n			
n	L^{∞} -error	order	L^{∞} -error	order	compression rate	
1	1.513187e-1		1.513187e-1		1.000000	
2	3.599605e-2	2.0717	3.599605e-2	2.0717	1.000000	
3	7.499082e-3	2.2631	7.514074e-3	2.2602	0.885742	
4	1.849162e-3	2.0198	1.848971e-3	2.0229	0.448761	
5	4.650610e-4	1.9914	4.737033e-4	1.9647	0.169880	

Strategy 3: $\epsilon_{i'i} = \nu \min\{d_{i'}, d_i\}$

Table 4.7: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 3 with $\nu = 1$).

Strategy 3: $\epsilon_{i'i} = \nu \min\{d_{i'}, d_i\}$

		$\nu = 1$			$\nu = 0.8$	
n	L_{∞} error	order	comp. rate	L_{∞} error	order	comp. rate
1	1.513187e-1		1.0000	1.513187e-1		1.0000
2	3.599605e-2	2.0717	1.0000	3.599605e-2	2.0717	1.0000
3	7.514074e-3	2.2602	0.885742	7.406683e-3	2.2809	0.674805
4	1.848971e-3	2.0229	0.448761	1.879626e-3	1.9784	0.301544
5	4.737033e-4	1.9647	0.169880	6.657693e-4	1.4974	0.111790

Table 4.8: Comparison of solutions obtained from strategy 3 with different ν

Strategy	2:	$\epsilon_{i'i} =$	$\nu \max$	$\{d_{i'},$	d_i
Strategy	4.	$c_{i'i}$ —	v man	(u_i)	ω_l J

		$\nu = 1$		$\nu = 0.8$			
n	L_{∞} error	order	comp. rate	L_{∞} error	order	comp. rate	
1	1.513187e-1		1.000000	1.513187e-1		1.000000	
2	3.599605e-2	2.0717	1.000000	3.599605e-2	2.0717	1.000000	
3	7.501118e-3	2.2627	0.894531	7.494318e-3	2.2640	0.753906	
4	1.847625e-3	2.0214	0.520172	1.850194e-3	2.0181	0.407013	
5	4.650882e-4	1.9901	0.227601	4.689467e-4	1.9802	0.174738	

Table 4.9: Comparison of solutions obtained from strategy 2 with different ν

		$\nu = 0.6$		$\nu = 0.4$			
n	L_{∞} error	order	comp. rate	L_{∞} error	order	comp. rate	
1	1.513187e-1		1.0000	1.513187e-1		1.0000	
2	3.599605e-2	2.0717	1.0000	3.599605e-2	2.0717	1.0000	
3	7.510530e-3	2.2609	0.712158	7.409970e-3	2.2803	0.641846	
4	1.855561e-3	2.0171	0.360596	1.887554e-3	1.9729	0.297562	
5	4.784063e-4	1.9555	0.148113	5.564827e-4	1.7621	0.118184	

Strategy 2: $\epsilon_{i'i} = \nu \max\{d_{i'}, d_i\}$

Table 4.10: Comparison of solutions obtained from strategy 2 with different ν

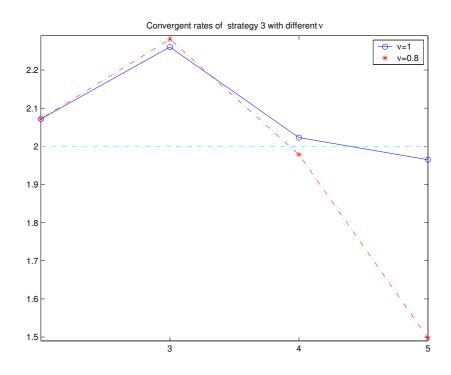


Figure 4.10: Convergent rates of strategy 3 with different ν

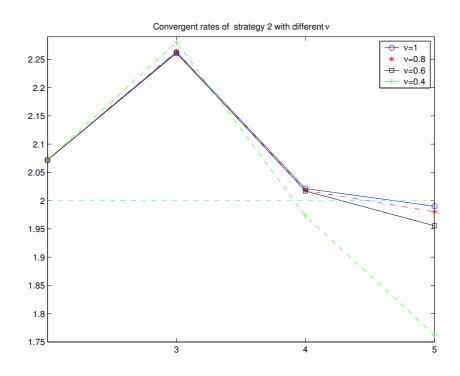


Figure 4.11: Convergent rates of strategy 2 with different ν

is relatively large, we are forced to use the sparse storage scheme to store the matrix (On our computer cluster.math.wvu.edu, with AMD Athlon(tm) MP Processor 1600+ and 1 Gigabyte of RAM, if serial codes are used, we have to use sparse storage when n > 5). For comparison of the compressed matrix $\tilde{\mathbf{K}}_n$ with the full matrix \mathbf{K}_n , in Fig. 4.15, we plot both $\tilde{\mathbf{K}}_3$ and \mathbf{K}_3 . Note in Fig. 4.15, for viewing effect, we use the absolute values of entries of the matrix $\tilde{\mathbf{K}}_3$ and \mathbf{K}_3 .

Next several theorems claim that if the constructed wavelets and collocation functionals on the unit triangle are employed to solve (2.2) on Ω , then similar results as in [CMX3] continue to hold.

Theorem 4.13 If the constructed wavelets and collocation functionals on the unit triangle in Chapter 3 are employed to solve (2.2) on the unit triangle Ω , and if the truncation parameters $\epsilon_{i'i}^n$, $i, i' \in \mathbb{Z}_{n+1}$ are chosen as (4.64) then there exists a positive constant c and a positive integer M such that when $n \geq M$ and for $x \in \mathbb{F}_n$,

$$\|(\mathcal{I} - \mathcal{K}_n)x\|_{\infty} \ge c \|x\|_{\infty},$$

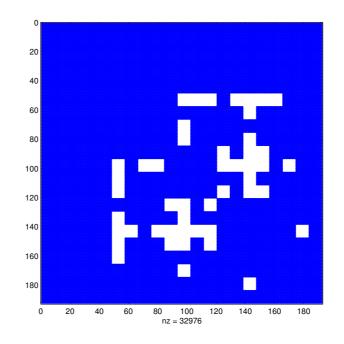


Figure 4.12: The sparse matrix \mathbf{K}_3 after compression

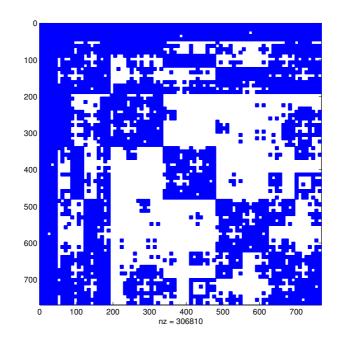


Figure 4.13: The sparse matrix \mathbf{K}_4 after compression

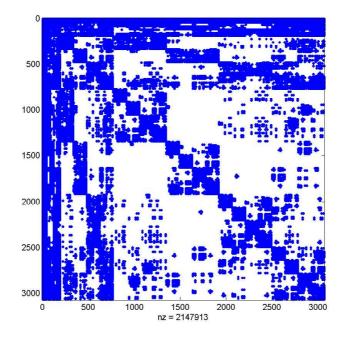


Figure 4.14: The sparse matrix \mathbf{K}_5 after compression

i.e., the compressed scheme (4.64) is stable, where, $\tilde{\mathcal{K}}_n : \mathbb{F}_n \mapsto \mathbb{F}_n$ is a linear operator relative to the basis $\{w_{ij} : (i,j) \in \mathbb{U}_n\}$ with $\mathbf{E}_n^{-1} \tilde{\mathbf{K}}_n$ as its matrix representation.

Proof: Since the constructed wavelet basis and collocation functionals have properties Lemma 3.1 through Lemma 3.7 and the kernel considered satisfying (4.23), the theorem follows from THEOREM 4.3 of [CMX3]. \Box

In Particular, this theorem allows for $n \ge M$,

$$(\mathcal{I} - \tilde{\mathcal{K}}_n)\tilde{u}_n = \mathcal{P}_n f \tag{4.68}$$

has a unique solution given by $\tilde{u}_n = \sum_{(i,j) \in \mathbb{U}_n} \tilde{u}_{ij} w_{ij}$.

Theorem 4.14 Use the constructed wavelets and collocation functionals on the unit triangle in Chapter 3 to solve (2.2) on Ω . Let $\epsilon_{i'i}^n$, $i, i' \in \mathbb{Z}_{n+1}$, be chosen as in (4.64). Then there exists a positive constant c and a positive integer M such that for all $n \geq M$,

$$||u - \tilde{u}_n||_{\infty} \le cf(n)^{-k/2} (\log f(n)) ||u||_{k,\infty}.$$

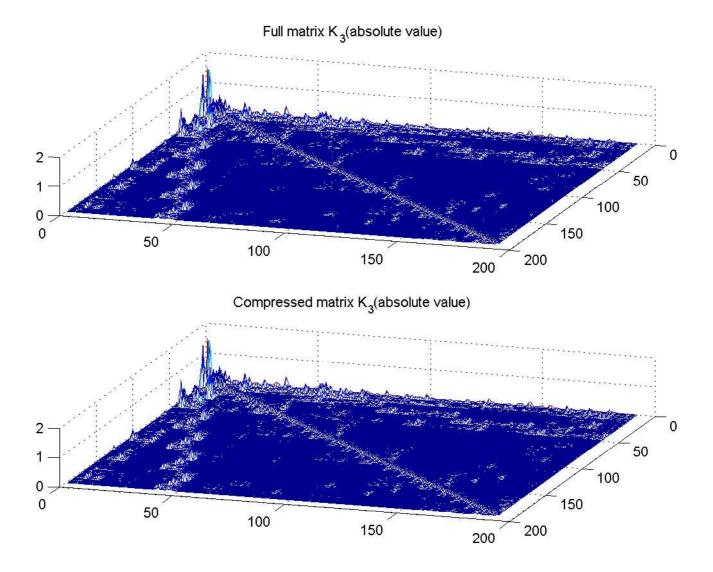


Figure 4.15: The Full matrix \mathbf{K}_3 and the compressed matrix $\tilde{\mathbf{K}}_3$

Proof: Since the constructed wavelet basis and collocation functionals have properties Lemma 3.1 through Lemma 3.8 and the kernel considered satisfying (4.23), the theorem follows from THEOREM 4.4 of [CMX3]. \Box

Theorem 4.15 If the conditions in THEOREM 4.13 hold, then there exists a positive constant c such that the condition number of the matrix $\tilde{\mathbf{A}}_n := \mathbf{E}_n - \tilde{\mathbf{K}}_n$ satisfies

$$\operatorname{cond}_{\infty}(\tilde{\mathbf{A}}_n) \le c \log^2(f(n)),$$

where $\operatorname{cond}_{\infty}(\mathbf{A})$ denotes the condition number of a matrix \mathbf{A} in ℓ^{∞} matrix norm.

Proof: Since the constructed wavelet basis and collocation functionals have properties Lemma 3.1 through Lemma 3.7 and the kernel considered satisfying (4.23), the theorem follows from THEOREM 4.5 of [CMX3]. \Box

Theorem 4.16 Use the constructed wavelets and collocation functionals on the unit triangle in Chapter 3 to solve (2.2) on Ω . Let the truncation parameters $\epsilon_{i'i}^n$, $i', i \in \mathbb{Z}_{n+1}$, be chosen as in (4.64). Then

$$\mathcal{N}(\mathbf{A}_n) = \mathcal{O}(f(n)\log f(n)),$$

Proof: Since the constructed wavelet basis and collocation functionals have property Lemma 3.1 and property Lemma 3.4, the theorem follows from THEOREM 4.6 of [CMX3]. \Box

4.6 Block Truncation Schemes

We describe in this section a block truncation scheme. This idea is also studied by [CWX1] in their study of 1-D wavelet collocation methods. By this scheme, the actual geometric distance dist $(\hat{S}_{i'j'}, S_{ij})$ is not needed to determine if an element $K_{i'j',ij}$, $(i', j'), (i, j) \in \mathbb{U}_n$ needs to be truncated. We first note

$$S_{ij} = \phi_{\mathbf{e}}(\Omega), \tag{4.69}$$

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and

$$\hat{S}_{i'j'} = \phi_{\mathbf{e}'}(\Omega). \tag{4.70}$$

 $\hat{S}_{i'j'}$ and S_{ij} are both right triangles transformed from the original unit triangle Ω . For S_{ij} , denote its vertex of right angle by $P_0(x_0, y_0)$, and counterclockwisely, denote the other two vertices by $P_1(x_1, y_1)$ and $P_2(x_2, y_2)$ (see Fig. 4.4). It is not difficult to find the centroid of the S_{ij} is at

$$C = \left(x_0 + \frac{x_1 - x_0}{3}, \ y_0 + \frac{x_1 - x_0}{3}\right) := (C_x, C_y).$$
(4.71)

We assign a multi-index $\mathbf{q} := (q_1, q_2)$ to S_{ij} , and \mathbf{q} is determined by

$$q_1 = \left\lfloor \frac{C_x}{\mu^{(i-1)/2}} \right\rfloor$$
 and $q_2 = \left\lfloor \frac{C_y}{\mu^{(i-1)/2}} \right\rfloor$. (4.72)

This can be written as

$$\mathbf{q} = \gamma(\mathbf{e}) \tag{4.73}$$

where γ is a function defined by $\gamma : \mathbb{Z}_{\mu}^{i-1} \mapsto \mathbb{Z}_{\mu^{(i-1)/2}}^2$.

If $i \ge i'$, let $\mathbf{e} = (e_0 e_1 \dots e_{i-2})$ and $\mathbf{e}' = (e'_0 e'_1 \dots e'_{i'-2})$. Denote $\mathbf{e}_c = (e_0 e_1 \dots e_{i'-2})$. We then have

$$\mathbf{q}' = (q_1', q_2') = \gamma(\mathbf{e}'),$$
(4.74)

and

$$\mathbf{q} = (q_1, q_2) = \gamma(\mathbf{e}_c). \tag{4.75}$$

If i < i', denote $\mathbf{e}'_{c} = (e'_{0}e'_{1} \dots e'_{i-2})$. We have

$$\mathbf{q}' = (q_1', q_2') = \gamma(\mathbf{e}_c'), \tag{4.76}$$

and

$$\mathbf{q} = (q_1, q_2) = \gamma(\mathbf{e}). \tag{4.77}$$

Now define a block $\mathbf{K}_{\mathbf{q'q}}^{i'i}$ by

$$\mathbf{K}_{\mathbf{q'q}}^{i'i} := \{ K_{i'j',ij} : \mathbf{q'}, \mathbf{q} \text{ are determined by } (4.74) - (4.77) \}.$$
(4.78)

We have, in both cases, that

$$\mathbf{K}_{i'i} = \left[\mathbf{K}_{\mathbf{q'q}}^{i'i} : \mathbf{q'}, \mathbf{q} \in \mathbb{Z}_{\mu^{(i_0-1)/2}}^2 \right],$$
(4.79)

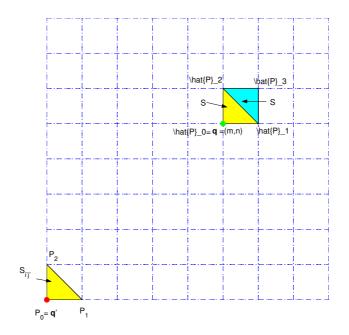


Figure 4.16: Distance of two triangles $S_{i'j'}$ and S

where $i_0 := \min\{i', i\}$. Figure 4.16 shows the idea of the block truncation scheme. If the support $\hat{S}_{i'j'}$ of the collocation functional $\ell_{i'j'}$ is in a finer level comparing with the support S_{ij} of the wavelet function w_{ij} , then the support of the same level as that of the wavelet containing $\hat{S}_{i'j'}$ is considered when \mathbf{q}' is assigned. Similar idea is applied if the support S_{ij} of the wavelet w_{ij} is in a finer level. Denote $S = \phi_{\mathbf{e}_c}(\Omega)$ if $i \ge i'$ and $S = \phi_{\mathbf{e}'_c}(\Omega)$ if i < i'. We have following lemma:

Lemma 4.17 Assume that $K_{i'j',ij}$ is an entry of $\mathbf{K}_{\mathbf{q'q}}^{i'i}$. Then, for $i \geq i'$,

$$\frac{|\mathbf{q} - \mathbf{q}'|}{\mu^{(i'-1)/2}} \ge \operatorname{dist}(\hat{S}_{i'j'}, S) \ge \frac{|\mathbf{q} - \mathbf{q}'| - \sqrt{2}}{\mu^{(i'-1)/2}},\tag{4.80}$$

and for i < i',

$$\frac{|\mathbf{q} - \mathbf{q}'|}{\mu^{(i-1)/2}} \ge \operatorname{dist}(S_{ij}, S) \ge \frac{|\mathbf{q} - \mathbf{q}'| - \sqrt{2}}{\mu^{(i-1)/2}}.$$
(4.81)

Proof: We present a proof for the case $i \ge i'$. The proof for the case i < i' is similar. Without loss of generality, assume vertices of $\hat{S}_{i'j'}$ are P_i , $i \in \mathbb{Z}_3$ with $P_0 = (0,0)$ (see Figure 4.16). Note $\hat{S}_{i'j'}$ is a right triangle with two equal sides. Each side is of length 1 unit, where the unit is $\frac{1}{\mu^{(i'-1)/2}}$. Following we only use the number of units to describe the length. With this assumption, we have $\mathbf{q}' = (0,0)$. Assume $\mathbf{q} = (m,n)$, where m, n are integers. Define $\hat{P}_0 = \mathbf{q}$, $\hat{P}_1 = (m+1,n)$, $\hat{P}_2 = (m,n+1)$, and $\hat{P}_3 = (m+1,n+1)$. Note S is either triangle $\hat{P}_0\hat{P}_1\hat{P}_2$ or triangle $\hat{P}_3\hat{P}_2\hat{P}_1$. We use $|P_i - P_j|$ to denote the distance between points P_i and P_j . Observed that

$$\operatorname{dist}(S, S_{i'j'}) \le \min\{|P_1 - \hat{P}_1|, |P_2 - \hat{P}_1|, |P_1 - \hat{P}_2|, |P_2 - \hat{P}_2|\}.$$
(4.82)

Now, $|\mathbf{q} - \mathbf{q}'| = \sqrt{m^2 + n^2}$, $|P_1 - \hat{P}_1| = \sqrt{(m)^2 + (n)^2} = |P_2 - \hat{P}_2|$, $|P_1 - \hat{P}_2| = \sqrt{(m-1)^2 + (n+1)^2}$; $|P_2 - \hat{P}_1| = \sqrt{(m+1)^2 + (n-1)^2}$.

Combining with (4.82), the first inequality follows. We now turn to prove the second inequality. If $\hat{S}_{i'j'}$ and S are not in the same row or column (see Figure 4.16), we have

dist
$$(S_{i'j'}, S) \ge \sqrt{(m-1)^2 + (n-1)^2} \ge \sqrt{m^2 + n^2} - \sqrt{2}$$

On the other hand, when $\hat{S}_{i'j'}$ and S are in the same row or column, by the triangle inequality, we have

$$\operatorname{dist}(\hat{S}_{i'j'}, S) + 1 \ge |\mathbf{q} - \mathbf{q}'|,$$

i.e.,

$$\operatorname{dist}(\hat{S}_{i'j'}, S) \ge |\mathbf{q} - \mathbf{q}'| - 1 \ge |\mathbf{q} - \mathbf{q}'| - \sqrt{2}$$

This completes the proof.

Lemma 4.18 Assume that $K_{i'j',ij}$ is an entry of $\mathbf{K}_{\mathbf{q'q}}^{i'i}$. Then

$$\frac{|\mathbf{q} - \mathbf{q}'| - \sqrt{2}}{\mu^{(i_0 - 1)/2}} \le \operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) < \frac{|\mathbf{q} - \mathbf{q}'| + \sqrt{2}}{\mu^{(i_0 - 1)/2}}$$
(4.83)

Proof: We present a proof for the case $i \ge i'$. The proof for the case i < i' is similar. First note that

$$S_{i'j'} = \phi_{\mathbf{e}'}(\Omega) \quad \text{and} \quad S_{ij} = \phi_{\mathbf{e}}(\Omega)$$

$$(4.84)$$

by (4.75) we have

$$\mathbf{q} = \gamma(\mathbf{e}_c), \quad \text{with } \mathbf{e}_c = (e_0, \dots, e_{i'-2}). \tag{4.85}$$

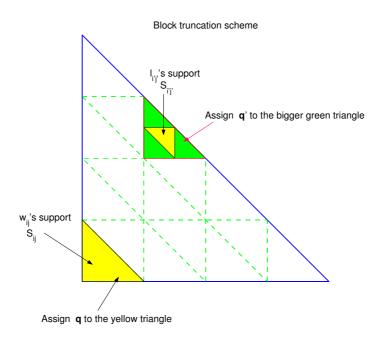


Figure 4.17: Block truncation scheme

Observe that

$$S_{ij} \subset S := \phi_{\mathbf{e}_c}(\Omega). \tag{4.86}$$

For $\mathbf{q} \neq \mathbf{q}'$, by Lemma 4.17

$$\frac{|\mathbf{q} - \mathbf{q}'|}{\mu^{(i'-1)/2}} \ge \operatorname{dist}(\hat{S}_{i'j'}, S) \ge \frac{|\mathbf{q} - \mathbf{q}'| - \sqrt{2}}{\mu^{(i'-1)/2}}.$$
(4.87)

This yields

$$\operatorname{dist}(\hat{S}_{i'j'}, S) \le \operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) < \operatorname{dist}(\hat{S}_{i'j'}, S) + \frac{\sqrt{2}}{\mu^{(i'-1)/2}}.$$
(4.88)

We now are ready to present the *block* truncation scheme. Define $\mathbf{p} := [p_{i'i} : p_{i'i} > 0, i', i \in \mathbb{Z}_{n+1}]$ and let

$$\tilde{\mathbf{K}}_{i'i} = \left[\mathbf{K}_{\mathbf{q'q}}^{i'i}(\mathbf{p}) : \mathbf{q'}, \mathbf{q} \in \mathbb{Z}_{\mu^{(i_0-1)/2}}^2 \right]$$
(4.89)

with

$$\mathbf{K}_{\mathbf{q'q}}^{i'i}(\mathbf{p}) = \begin{cases} \mathbf{K}_{\mathbf{q'q}}^{i'i}, & |\mathbf{q} - \mathbf{q'}| \le p_{i'i} \\ 0, & otherwise. \end{cases}$$
(4.90)

Next theorem shows that solutions of (4.68) by this block truncation scheme has the same order of convergence and computational complexity as those by adopting the theoretical truncation scheme (4.64), provided that the truncation parameters **p** is properly chosen.

Theorem 4.19 Assume that $u \in W^{k,\infty}(\Omega)$. For some a > 0 and $\nu' > 0$, let

$$p_{i'i} := \max\{a\mu^{[b'(n-i')+(i'-i-1)]/2}, \nu'(\mu^{(-i+i')/2}+1)\} - \sqrt{2}, \quad \text{for } i' < i$$
(4.91)

of

$$p_{i'i} := \max\{a\mu^{[b'(n-i')-1]/2}, \nu'(\mu^{(i-i')/2}+1)\} - \sqrt{2}, \quad \text{for } i' \ge i.$$
(4.92)

Assume scheme (4.89), then there exists a positive constant c such that

$$\|u - \tilde{u}_n\|_{\infty} \le cf(n)^{-k/2} \log f(n) \|u\|_{k,\infty}$$
(4.93)

and

$$\mathcal{N}(\mathbf{A}_n) = \mathcal{O}\left(f(n)\log f(n)\right),\tag{4.94}$$

where $\mathcal{N}(\tilde{\mathbf{A}}_n)$ denotes the number of nonzero entries in matrix $\tilde{\mathbf{A}}_n$.

Proof: As proved in [CMX3] if the truncation parameter $\epsilon_{i'i}$ is chosen by (4.64), and scheme (4.63) is applied, then estimate (4.93) holds. It suffices to show that if the block truncation scheme (4.89) is assumed, scheme (4.63) is preserved. Namely, we need to show for appropriate \mathbf{p} , if $|\mathbf{q} - \mathbf{q}'| \leq p_{i'i}$, then

$$\operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) \le \epsilon_{i'i}.\tag{4.95}$$

From (4.78) we have for each pair $(i', j'), (i, j) \in \mathbb{U}_n$, there exists $\mathbf{q}', \mathbf{q} \in \mathbb{Z}^2_{\mu^{(i_0-1)/2}}$ such that $K_{i'j',ij}$ is an entry of the block $\mathbf{K}_{\mathbf{qq}'}^{i'i}$. If

$$|\mathbf{q} - \mathbf{q}'| < p_{i'i}$$

in view of (4.91), (4.92) and (4.64), we have

$$\frac{|\mathbf{q} - \mathbf{q}'| + \sqrt{2}}{\mu^{(i_0 - 1)/2}} \le \epsilon_{i'i}.$$

It follows from Lemma 4.18 that

$$\operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) \le \epsilon_{i'i}. \tag{4.96}$$

According to the block truncation strategy (4.89), we have that

$$\tilde{\mathbf{K}}^{i'i} = [\mathbf{K}^{i'i}_{\mathbf{q}'\mathbf{q}}(\mathbf{p}) : |\mathbf{q} - \mathbf{q}'| \le p_{i'i}]$$
(4.97)

which guarantees the scheme (4.63) and hence (4.93) follows. It remains to prove (4.94). To this end we set

$$\Lambda_1 := \{ (i', j'), (i, j) \in \mathbb{U}_n : |\mathbf{q}' - \mathbf{q}| \le p_{i'i} \},$$
(4.98)

$$\Lambda_2 := \{ (i', j'), (i, j) \in \mathbb{U}_n : \operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) \le \epsilon_{i'i} \},$$
(4.99)

$$\Lambda_3 := \{ (i', j'), (i, j) \in \mathbb{U}_n : \epsilon_{i'i} < \operatorname{dist}(\hat{S}_{i'j'}, S_{ij}) \text{ and } |\mathbf{q}' - \mathbf{q}| \le p_{i'i} \}.$$
(4.100)

We estimate the cardinality $\operatorname{card}(\Omega_1)$. we have seen inequality $|\mathbf{q}' - \mathbf{q}| \leq p_{i'i}$ implies that $\operatorname{dist}(S_{i'j'}, S_{ij}) \leq \epsilon_{i'i}$. Thus, we have that

$$\operatorname{card}(\Lambda_1) = \operatorname{card}(\Lambda_2) + \operatorname{card}(\Lambda_3).$$

It is proved in [CMX3] that

$$\operatorname{card}(\Lambda_2) = \mathcal{O}(f(n)\log f(n)). \tag{4.101}$$

Now for any fixed $i', i \in \mathbb{Z}_{n+1}$, there are at most $c\mu^{i_0-1}$ sub-blocks $\mathbf{K}_{\mathbf{qq}'}^{i'i}$ in $\mathbf{K}_{i'i}$ for some positive constant c, satisfying that both $\epsilon_{i'i} < \operatorname{dist}(\hat{S}_{i'j'}, S_{ij})$ and $|\mathbf{q}' - \mathbf{q}| \le p_{i'i}$. While each sub-block $\mathbf{K}_{\mathbf{q'q}}^{i'i}$ has at most $2r^2\mu^{|i-i'|}$ entries. Therefore

$$\operatorname{card}(\Lambda_3) \le 4cr^2 \sum_{i \in \mathbb{Z}_{n+1}} \sum_{i' \in \mathbb{Z}_{i+1}} \mu^{|i-i'|+i_0-1} = \mathcal{O}(f(n)\log f(n)).$$

This with (4.101) establishes (4.94).

For any given triangle $\hat{S}_{i'j'}$ associated with index \mathbf{q}' , the block truncation scheme is expected to catch any triangle S associated with index \mathbf{q} surrounding $\hat{S}_{i'j'}$. On the other hand, we don't want too many triangles are caught by the scheme. A simple analysis shows that for triangles which are surrounding $\hat{S}_{i'j'}$, $|\mathbf{q}' - \mathbf{q}| \leq \sqrt{2}$, hence $\sqrt{2} < p_{i'i} \leq \sqrt{3}$, $i', i \in \mathbb{Z}_{n+1}$

would be a very good choice. Following examples demonstrates this.

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For the compression rates of the block truncation scheme with $p_{i'i} = \sqrt{3}$ please see Table 4.4. For comparing compression rates with those of other schemes, please see Figure 4.8. Note the block truncation scheme is not contained in any of the schemes discussed in the previous section. While for distance computing scheme: strategy 1 contains strategy 2 and in turn strategy 2 contains strategy 3. When level n < 5, the block truncation scheme has maximum compression rate, yet, the convergence rate is very satisfactory. When n > 5, the compression rate of block truncation scheme is between that of strategy 2 and that of strategy 3, nevertheless, block truncation scheme remains to give good solution. This suggests that, even compared to strategy 2, block truncation scheme can give more or less the same good solution but with less computational effort. In view of Figure 4.9 and the computing results in Table 6.1 and Table 6.2 in solving the integral equations (6.1), when n = 8, the block truncation scheme is the best one among all the *practical* compression strategies we discussed so far.

Example 4.20 In this example, we solve the integral equation (6.1). To test if the proposed block truncation strategy is good enough, all the elements are computed by analytical method and the linear system of equations are solved by standard Gaussian elimination method. The results are tabulated in Table 4.11. We can see that the compressed solutions by block truncation parameter cation scheme follow very closely the uncompressed solution. The block truncation parameter $p_{i'i}$ is set as $\sqrt{3}$. The convergent rates for different levels are plotted in Figure 4.9.

4.7 Quadrature Error Control

After discretizing the integral equation (2.2) with compression applied, we obtain a linear system

$$\mathbf{E}_n - \tilde{\mathbf{K}}_n = \mathbf{f}_n. \tag{4.102}$$

Since the nonzero entries of the coefficient matrix \mathbf{K}_n (obtained after compression) are usually obtained by quadrature, we actually solve a perturbed version of the linear system (4.102),

	full collocation solution u_n		compressed solution \tilde{u}_n			
n	L^{∞} -error	order	L^{∞} -error	order	compression rate	
1	1.513187e-1		1.513187e-1		1.000000	
2	3.599605e-2	2.0717	3.599605e-2	2.0717	1.000000	
3	7.499082e-3	2.2631	7.493692e-3	2.2641	0.758301	
4	1.849162e-3	2.0198	1.851165e-3	2.0172	0.416626	
5	4.650610e-4	1.9914	4.683404e-4	1.9828	0.181021	

Block truncation scheme with $p_{i'i} = \sqrt{3}$

Table 4.11: Comparison of solutions obtained by the full collocation method and by the compressed method (with block truncation scheme).

say,

$$(\mathbf{E}_{\mathbf{n}} - \tilde{\mathbf{K}}_n)\tilde{\tilde{\mathbf{u}}}_n = \mathbf{f}_n, \qquad (4.103)$$

where

$$\tilde{\tilde{\mathbf{K}}}_n = (\tilde{\tilde{K}}_{i'j',ij})$$
 with $\tilde{\tilde{K}}_{i'j',ij} = \tilde{K}_{i'j',ij} + \delta_{i'j',ij}$

and $\delta_{i'j',ij}$ denotes the quadrature error for computing the integral $K_{i'j',ij}$ (or equivalently, $\tilde{K}_{i'j',ij}$). We wish to control quadrature errors, such that they will not affect the convergent rate of the approximation solutions. In the following we establish a rule for the quadrature accuracy so that the solution $\tilde{\tilde{u}}_n$ obtained still satisfies the same order of error estimate as stated in Theorem 4.14. Let $\delta_{i'i}$ be the quadrature error for the block matrix $\tilde{\mathbf{K}}_{i'i}$, namely

$$\delta_{i'i} = \max_{j',j} |K_{i'j',ij} - \tilde{\tilde{K}}_{i'j',ij}|, \qquad j \in \mathbb{Z}_{w(i)}, \ j' \in \mathbb{Z}_{w(i')}.$$

Lemma 4.21 Suppose that for a given constant $\nu > 1$ and $0 < \sigma' < 2 - \sigma$, the quadrature rule have an accuracy such that

$$\delta_{i'i} \le c\epsilon_{i'i}^{-\eta} d_i^{2+k} d_{i'}^k, \quad i', i \in \mathbb{Z}_{n+1},$$
(4.104)

for some constant c. Then there exist a constant c_1 whenever $\epsilon_{i'i} \ge r(d_i + d_{i'})$ then

$$\|\mathbf{K}_{i'i} - \tilde{\mathbf{K}}_{i'i}\|_{\infty} \le c_1 \epsilon_{i'i}^{-\eta} (d_i d_{i'})^k, \quad i', i \in \mathbb{Z}_{n+1},$$
(4.105)

where $\epsilon_{i'i}$ is the truncation parameter for the block matrix $\mathbf{K}_{i'i}$ and $\eta := 2k - \sigma'$.

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 $\it Proof:$ We first note

$$\|\mathbf{K}_{i'i} - ilde{\mathbf{K}}_{i'i}\|_{\infty} \leq \|\mathbf{K}_{i'i} - ilde{\mathbf{K}}_{i'i}\|_{\infty} + \| ilde{\mathbf{K}}_{i'i} - ilde{\mathbf{K}}_{i'i}\|_{\infty}$$

From Lemma 3.2 of [CMX3] we have

$$\|\mathbf{K}_{i'i} - \tilde{\mathbf{K}}_{i'i}\|_{\infty} \le c\epsilon_{i'i}^{-\eta} (d_{i'}d_i)^k \qquad i, i' \in \mathbb{Z}_{n+1}$$

It then follows from (4.104) that

$$\begin{split} \|\tilde{\mathbf{K}}_{i'i} - \tilde{\tilde{\mathbf{K}}}_{i'i}\|_{\infty} &= \max_{j' \in \mathbb{Z}_{w(i')}} \sum_{j \in \mathbb{Z}_{w(i)} \setminus \mathbb{Z}_{i'j'}(\epsilon)} |K_{i'j',ij} - \tilde{\tilde{K}}_{i'j',ij}| \\ &\leq \max_{j' \in \mathbb{Z}_{w(i')}} \sum_{j \in \mathbb{Z}_{w(i)}} |K_{i'j',ij} - \tilde{\tilde{K}}_{i'j',ij}| \leq cw(i)\epsilon_{i'i}^{-\eta}(d_i d_{i'})^k d_i^2 \\ &\leq c\epsilon_{i'i}^{-\eta}(d_i d_{i'})^k, \end{split}$$

where

$$\mathbb{Z}_{i'j'}(\epsilon) := \{ j : j \in \mathbb{Z}_{w(i)}, \operatorname{dist}(S_{ij}, \hat{S}_{i'j'}) > \epsilon_{i'i} \}.$$

The last inequality is obtained by using

$$w(i) \le c\mu^i \tag{4.106}$$

and

$$d_i = c\mu^{-i/2}. (4.107)$$

This completes the proof.

Corrolary 4.22 If the truncation parameter $\epsilon_{i'i}$ is chosen to be $c(d_i + d_{i'})$ for some constant c, then (4.105) is satisfied by choosing

$$\delta_{i'i} \le c(d_i)^{k+2} d_{i'}^k \tag{4.108}$$

or equivalently by choosing

$$\delta_{i'i} \le c\mu^{-k(\frac{i}{2} + \frac{i'}{2}) - i}.$$
(4.109)

Proof: From (4.107) and $d_i \leq c, i \in \mathbb{N}_0$, we obtain

$$c2^{-\eta} \le (d_i + d_{i'})^{-\eta}.$$

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Now

$$c(d_i d_{i'})^k d_i^2 = c' 2^{-\eta} (d_i d_{i'})^k d_i^2 \le c(d_i + d_{i'})^{-\eta} (d_i d_{i'})^k d_i^2 = c \epsilon_{i'i}^{-\eta} (d_i d_{i'})^k d_i^2$$

Thus (4.104) is satisfied, and (4.105) follows. (4.109) follows by directly using (4.108) and (4.107).

Let $m_{i'i}$ be the number used in the quadrature rule (see (4.36)) to compute entries of the block matrix $\tilde{\tilde{\mathbf{K}}}_{i'i}$. Assume that m_{00} is chosen. If the quadrature rule (4.36) has error estimate $O(m^{-2k})$, then $m_{i'i}$ can be chosen as

$$m_{i'i} = cm_{00}(\epsilon_{i'i})^{\eta/2k} (d_i d_{i'})^{-1/2} d_i^{-\frac{1}{k}}.$$
(4.110)

In particular, if $\epsilon_{i'i} = c(d_i + d_{i'})$, we can choose

$$m_{i'i} = cm_{00} (d_i d_{i'})^{-1/2} (d_i)^{-1/k}.$$
(4.111)

Theorem 4.23 Assume that the quadrature rule used to compute entries of $\tilde{\mathbf{K}}_{\mathbf{n}}$ has error estimate $O(m^{-2k})$. Choose m_{00} large enough such that (4.104) is satisfied for i' = i = 0and identify other $m_{i'i}$ by (4.110) (in the case of $\epsilon_{i'i} = c(d_{i'} + d_i)$, (4.111) is used), and the truncation parameter $\epsilon_{i'i}$, $i', i \in \mathbb{Z}_{n+1}$ are chosen according to (4.64). Then there exists a positive constant c and a positive integer N such that for all n > N,

$$\|u - \tilde{\tilde{u}}_n\|_{\infty} \le cf(n)^{-k/2} \log f(n) \|u\|_{\infty},$$

where, $\tilde{\tilde{u}}_n = \sum_{(i,j) \in \mathbb{U}_n} \tilde{\tilde{u}}_{ij} w_{ij}$, with $\tilde{\tilde{\mathbf{u}}}_n = [\tilde{\tilde{u}}_{ij}, (i,j) \in \mathbb{U}_n].$

Proof: Since the quadrature rule has error estimate $O(m^{-2k})$, if m_{00} is large enough to meet (4.104), there holds

$$cm_{00}^{-2k} \le \epsilon_{00}^{-\eta} d_0^{2k+2}.$$

By (4.110) we have

$$m_{i'i}^{-2k} = c(m_{00})^{-2k} \left(\epsilon_{i'i}^{\eta/2k} (d_i d_{i'})^{-1/2} d_i^{-1/k} \right)^{-2k} \le c \epsilon_{i'i}^{-\eta} (d_i d_{i'})^k d_i^2.$$

Therefore (4.104) is satisfied. For the case when $\epsilon_{i'i} = c(d_i + d_{i'})$, use (4.111), (4.104) is satisfied. The remaining proof can follow that of [CMX3](see [CMX3] Lemma 4.1, Lemma 4.2 and Theorem 4.4).

We next discuss the quadrature error control from another perspective. Assuming both the coefficient matrix $\tilde{\mathbf{K}}_n$ and \mathbf{f}_n are obtained by numerical quadrature. In this case, we solve a perturbed version of the linear system (4.102),

$$(\mathbf{E}_n - \tilde{\mathbf{K}}_n)\tilde{\mathbf{u}}_n = \tilde{\mathbf{f}}_n, \qquad (4.112)$$

where

$$\tilde{\tilde{\mathbf{K}}}_n = (\tilde{\tilde{K}}_{i'j',ij}) \quad \text{with} \quad \tilde{\tilde{K}}_{i'j',ij} = \tilde{K}_{i'j',ij} + \delta_{i'j',ij},$$
$$\tilde{\tilde{\mathbf{f}}} = (\tilde{\tilde{f}}_{i'j'}) \quad \text{with} \quad \tilde{\tilde{f}}_{i'j'} = f_{i'j'} + \delta_{i'j'}.$$

 $\delta_{i'j',ij}$ and $\delta_{i'j'}$ denote the quadrature error for computing the double integral $K_{i'j',ij}$ and single integral $f_{i'j'}$, respectively. We desire the solution $\tilde{\tilde{u}}_n$ still satisfies the same order of error estimate as the compressed collocation method (see Theorem 4.14). Namely,

$$\|\tilde{u}_n - u\|_{\infty} = O(n\mu^{-kn/2}), \tag{4.113}$$

where, k is the total degree of piecewise polynomials in \mathbb{F}_n . Since w_{ij} , $(i, j) \in \mathbb{U}$ is uniformly bounded (Lemma 3.4), we have that

$$\|\tilde{u}_n - \tilde{\tilde{u}}_n\|_{\infty} \le C \|\tilde{\mathbf{u}}_n - \tilde{\tilde{\mathbf{u}}}_n\|_{\infty}.$$

Moreover, because $\tilde{\mathbf{u}}_n - \tilde{\tilde{\mathbf{u}}}_n$ solves the linear system

$$(\mathbf{E}_n - \tilde{\mathbf{K}}_n)(\tilde{\mathbf{u}}_n - \tilde{\tilde{\mathbf{u}}}_n) = \mathbf{f}_n - \tilde{\tilde{\mathbf{f}}}_n + (\tilde{\mathbf{K}}_n - \tilde{\tilde{\mathbf{K}}}_n)\tilde{\tilde{\mathbf{u}}}_n,$$

we conclude that

$$\|\tilde{u}_n - \tilde{\tilde{u}}_n\|_{\infty} \le C \|\tilde{\mathbf{u}}_n - \tilde{\tilde{\mathbf{u}}}_n\|_{\infty} = \mathcal{O}(\|\mathbf{f}_n - \tilde{\tilde{\mathbf{f}}}_n\|_{\infty}) + \mathcal{O}(\|\tilde{\mathbf{K}}_n - \tilde{\tilde{\mathbf{K}}}_n\|_{\infty}).$$

Now,

$$\|\mathbf{f}_n - \tilde{\mathbf{f}}_n\|_{\infty} = \max_{(i',j') \in \mathbb{U}_n} |\delta_{i'j'}|,$$

and

$$\|\tilde{\mathbf{K}}_n - \tilde{\tilde{\mathbf{K}}}_n\|_{\infty} \le \max_{(i',j') \in \mathbb{U}_n} \sum_{(i,j) \in \mathbb{U}_n} |\delta_{i'j',ij}| \le c\mu^n \max_{(i',j'),(i,j) \in \mathbb{U}_n} |\delta_{i'j',ij}|.$$

Thus, by letting

$$\delta^{(n)} := \max_{(i',j'),(i,j) \in \mathbb{U}_n} \{ |\delta_{i'j'}|, |\delta_{i'j',ij}| \},\$$

we have

$$\|\tilde{u}_n - \tilde{\tilde{u}}_n\|_{\infty} = \mathcal{O}(\mu^n \delta^{(n)})$$

In order for $\tilde{\tilde{u}}_n$ to be an approximation to u_n in the same order as \tilde{u}_n to u_n in (4.113), we need to require that the quadrature error $\delta^{(n)}$ be reduced in the order of

$$\delta^{(n)} \sim n\mu^{\frac{-kn}{2}-n}$$

as n increases. When apply the quadrature rule presented in Section 4.3, this requirement implies that the number of subdivisions, m_n , according to n need to be increased as

$$m_n \sim n^{-\frac{1}{2k}} \mu^{\frac{(k+2)n}{4k}}.$$
 (4.114)

That is,

$$\frac{m_{n+1}}{m_n} = \left(\frac{n}{n+1}\right)^{\frac{1}{2k}} \mu^{\frac{(k+2)}{4k}},\tag{4.115}$$

which implies that, as the level of resolution increases by 1, the number of subdivisions in the quadrature rule should be increased by approximately $\mu^{\frac{k+2}{4k}}$ times, in order to maintain the error order of the approximation solution of the collocation method as shown in (4.113). In our examples, k = 2, $\mu = 4$, hence this factor is approximately $4^{1/2} = 2.0$. This method to control the quadrature error is demonstrated in Example 4.25. When the level $i \ge 2$, the domain of integration is no longer the whole domain and getting smaller and smaller. If the element $f_{i'j'}$ is obtained also by numerical quadrature, then its domain of integration is always the whole original domain. In view of Table 4.1 and Table 4.2, in practical computing, we can expect to use less number of subdivisions to compute elements $K_{i'j',ij}$, for $i \ge 2$ to obtain the same order of convergence rate. This idea is demonstrated in Example 4.26. We summarize above discussion in the following theorem:

Theorem 4.24 If the number of subdivisions m_n used in the adaptive quadrature rule (4.36) is chosen as (4.114) for solving the perturbed equation (4.112), then

$$\|u - \tilde{\tilde{u}}_n\|_{\infty} \le cf(n)^{-k/2} \log f(n) \|u\|_{\infty}.$$

Example 4.25 In this example, we test different number of subdivisions in the quadrature rule for different level n when solving the integral equation (6.1). But for fixed n, the computing of all entries assumes the same number of subdivisions m_n for the quadrature rule.

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	without qua	drature	with quadrature			
n	L^{∞} -error	order	m_n	L^{∞} -error	order	
1	1.513187e-1		10	1.512309e-1		
2	3.599605e-2	2.0717	10	3.540331e-2	2.0948	
3	7.499082e-3	2.2631	10	7.526215e-3	2.2339	
4	1.849162e-3	2.0198	20	1.864510e-3	2.0131	
5	4.650610e-4	1.9914	40	4.657507e-4	2.0012	

Table 4.12: Comparison of uncompressed solutions, with and without numerical quadratures, using different m_n for different n.

To eliminate other affecting factors, the linear system of equations are solved by standard Gaussian elimination method. No compression is assumed. The results are tabulated in Table 4.12. We can see from Table 4.12 that the results by using the quadrature control rule (4.115) are very close to those obtained without quadrature rule.

Example 4.26 In this example, we test for fixed n, different number of subdivisions for different levels used in the quadrature rule when solving the integral equation (6.1). The linear system of equations are solved by standard Gaussian elimination method. No compression is assumed. The results are tabulated in Table 4.13. In Table 4.13, (m_1, m_2) means that when the level i, $i \in \mathbb{Z}_{n+1}$ corresponding to wavelet functions w_{ij} less than 2, the number of subdivisions used in quadrature is m_1 , and when $i \ge 2$, the number of subdivisions used in quadrature assumes m_2 . While for computing $f_{i'j'}$, the number of subdivisions used is always m_1 . We can see from results in Table 4.13 that the this quadrature control method also works very well too. But, the computational cost compared to that of Example 4.25 for quadrature has been dramatically saved.

4.8 Iteration Methods

In this section, we develop multilevel algorithms to solve the resulting linear system by making use of multilevel structures of the approximation spaces \mathbb{F}_n of X. In [CMX4] a two-

	without qua	drature	with quadrature			
n	L^{∞} -error	order	(m_1,m_2)	L^{∞} -error	order	
1	1.513187e-1		(10, 10)	1.512309e-1		
2	3.599605e-2	2.0717	(10, 10)	3.540331e-2	2.0948	
3	7.499082e-3	2.2631	(20, 10)	7.548485e-3	2.2296	
4	1.849162e-3	2.0198	(20, 10)	1.852722e-3	2.0265	
5	4.650610e-4	1.9914	(40, 20)	4.638874e-4	1.9978	

Table 4.13: Comparison of uncompressed solutions, with and without numerical quadratures, using different numbers of subdivisions for different levels.

level method is introduced and analyzed for solving operator equations in a Banach Space. [CWX1] also provides a two-level method called *augmentation method* utilizing the multilevel structure of the approximation space. In [FMX], iterative methods are developed for solving the linear system resulting from wavelet Galerkin discretization.

Remember that the wavelet collocation scheme for equation (2.2) is to seek $u_n \in \mathbb{X}$ (hence $u_n \in \mathbb{F}_n$) satisfying

$$(\mathcal{I} - \mathcal{P}_n \mathcal{K})u_n = \mathcal{P}_n f, \qquad (4.116)$$

or equivalently,

$$(\mathcal{I} - \mathcal{P}_n \mathcal{K} \mathcal{P}_n) u_n = \mathcal{P}_n f, \qquad (4.117)$$

where, \mathcal{P}_n is defined in (3.52) and we have

$$\mathbb{F}_n = \mathcal{P}_n \mathbb{X}$$

It is well known (cf. [At2]) by the compactness of \mathcal{K} and the pointwise convergence (see Lemma 3.7) of \mathcal{P}_n that, for any sufficient large n, there exists a constant C, such that

$$\|(\mathcal{I} - \mathcal{P}_n \mathcal{K})^{-1}\| \le C \tag{4.118}$$

and

$$\lim_{n \to \infty} \|\mathcal{K} - \mathcal{P}_n \mathcal{K}\| = \lim_{n \to \infty} \|\mathcal{K} - \mathcal{K} \mathcal{P}_n\| = 0.$$
(4.119)

Recall we have required the nestedness of the approximation spaces, namely

$$\mathbb{F}_n \subset \mathbb{F}_{n+1}, \ n \in \mathbb{N}_0, \tag{4.120}$$

and the space \mathbb{F}_n is decomposed into a direct sum of a family of multiscale spaces, i.e.,

$$\mathbb{F}_n = \mathbb{W}_0 \oplus \mathbb{W}_1 \oplus \ldots \oplus \mathbb{W}_n. \tag{4.121}$$

Note $\mathbb{F}_0 = \mathbb{W}_0$ and

$$\mathbb{F}_n = \mathbb{F}_{n-1} \oplus W_n, \quad n = 1, 2, \dots$$
(4.122)

Next lemma shows that

$$\mathcal{Q}_n := \mathcal{P}_n - \mathcal{P}_{n-1}, \quad n = 1, 2, \dots$$

are projections on X.

Lemma 4.27 Let \mathcal{P}_1 , \mathcal{P}_2 be projections on a Banach space X, Then $\mathcal{Q} = \mathcal{P}_2 - \mathcal{P}_1$ is a projection if and only if

$$\mathbb{Y}_1 \subset \mathbb{Y}_2,$$

where $\mathbb{Y}_j = \mathcal{P}_j, \ j = 1, 2.$

Proof: Assume $Q = P_2 - P_1$ is a projection, then

$$\mathcal{P}_2 - \mathcal{P}_1 = \mathcal{P}_2 + \mathcal{P}_1 - \mathcal{P}_2 \mathcal{P}_1 - \mathcal{P}_1 \mathcal{P}_2,$$

i.e,,

$$2\mathcal{P}_1 = \mathcal{P}_2\mathcal{P}_1 + \mathcal{P}_1\mathcal{P}_2.$$

Multiply above formula by \mathcal{P}_2 from left and right respectively, we have

$$\mathcal{P}_1\mathcal{P}_2=\mathcal{P}_2\mathcal{P}_1=\mathcal{P}_1.$$

Thus for $x \in \mathbb{X}$, $\mathcal{P}_1 x = \mathcal{P}_2 \mathcal{P}_1 x$, namely, $\mathbb{Y}_1 \subset \mathbb{Y}_2$. Conversely, let $\mathbb{Y}_1 \subset \mathbb{Y}_2$, for $x \in \mathbb{X}$, $\mathcal{P}_1 x \in \mathbb{Y}_2$, thus $\mathcal{P}_2 \mathcal{P}_1 x = \mathcal{P}_1 x$, hence, $\mathcal{P}_2 \mathcal{P}_1 = \mathcal{P}_1$. We then have $\mathcal{P}_1 \mathcal{P}_2 \mathcal{P}_1 = \mathcal{P}_1 \mathcal{P}_1$, which is equivalent to $(\mathcal{P}_1 \mathcal{P}_2 - \mathcal{P}_1)\mathcal{P}_1 = 0$, leading to $\mathcal{P}_1 \mathcal{P}_2 = \mathcal{P}_1$ in any case. This completes the proof.

From Lemma 4.27, we obtain

$$\mathbb{W}_n = \mathcal{Q}_n \mathbb{F}_n, \quad n = 1, 2, \dots$$

This yields, for $l, k \ge 0$,

$$\mathcal{P}_{k+\ell} = \mathcal{P}_k + \mathcal{Q}_{k+1} + \dots + \mathcal{Q}_{k+\ell}.$$

Observe that (4.122) yields, for $k \ge 0$, $\ell \ge 1$, and $n = k + \ell$,

$$\mathbb{F}_n = \mathbb{F}_k \oplus \mathbb{W}_{k+1} \oplus \dots \oplus \mathbb{W}_n. \tag{4.123}$$

For the convenience of description, let's identify $[f_0, g_1, \ldots, g_\ell]^T$ in $\mathbb{F}_k \times \mathbb{W}_{k+1} \times \cdots \times W_{k+\ell}$ with the vector

$$f_0 + g_1 + \dots + g_\ell \in \mathbb{F}_k \oplus \mathbb{W}_{k+1} \oplus \dots \oplus \mathbb{W}_{k+\ell}.$$

In particular, for $u_{k+\ell} \in \mathbb{F}_{k+\ell}$,

$$u_{k+\ell} = u_{k,0} + v_{k,1} + \dots + v_{k,\ell} = [u_{k,0}, v_{k,1}, \dots, v_{k,\ell}]^T,$$

where $u_{k,0} \in \mathbb{F}_k$ and $v_{k,i} \in \mathbb{W}_{k+i}$ $(1 \le i \le \ell)$. Now we define , for $n, m \ge 0$,

$$\mathcal{K}_n = \mathcal{P}_n \mathcal{K} \mathcal{P}_n, \quad \mathcal{K}_{n,m} = \mathcal{Q}_n \mathcal{K} \mathcal{Q}_m,$$

$$\mathcal{B}_{n,m} = \mathcal{P}_n \mathcal{K} \mathcal{Q}_m \text{ for } n < m, \text{ and } \mathcal{C}_{n,m} = \mathcal{Q}_n \mathcal{K} \mathcal{P}_m \text{ for } n > m.$$

With these notations, the operator $\mathcal{K}_{k+\ell}$ can be identified with

$$\mathcal{A}_{k,\ell} = \begin{bmatrix} \mathcal{K}_k & \mathcal{B}_{k,k+1} & \cdots & \mathcal{B}_{k,k+\ell} \\ \mathcal{C}_{k+1,k} & \mathcal{K}_{k+1,k+1} & \cdots & \mathcal{K}_{k,k+\ell} \\ \vdots & \vdots & \ddots & \vdots \\ \mathcal{C}_{k+\ell,k} & \mathcal{K}_{k+\ell,k+1} & \cdots & \mathcal{K}_{k+\ell,k+\ell} \end{bmatrix}.$$
(4.124)

This notation was first introduced by [CMX4]. Thus, equation (4.116) or (4.117) can be written into

$$\mathbf{u}_{k,\ell} - \mathcal{A}_{k,\ell} \mathbf{u}_{k,\ell} = \mathbf{f}_{k,\ell},\tag{4.125}$$

where $\mathbf{u}_{k,\ell} = [u_{k,0}, v_{k,1}, \dots, v_{k,\ell}]^T$ and $\mathbf{f}_{k,\ell} = [\mathcal{P}_k f, \mathcal{Q}_{k+1} f, \dots, \mathcal{Q}_{k+\ell} f]^T$. Let

$$\mathcal{U}_{k,\ell} = \begin{bmatrix} O & \mathcal{B}_{k,k+1} & \mathcal{B}_{k,k+2} & \cdots & \mathcal{B}_{k,k+\ell} \\ & O & \mathcal{K}_{k+1,k+2} & \cdots & \mathcal{K}_{k+1,k+\ell} \\ & \ddots & & \vdots \\ & & \ddots & \mathcal{K}_{k+\ell-1,k+\ell} \\ & & & O \end{bmatrix},$$

$$\mathcal{L}_{k,\ell} = \begin{bmatrix} O \\ \mathcal{C}_{k+1,k} & \mathcal{K}_{k+1,k+1} \\ \vdots & \vdots & \ddots \\ \mathcal{C}_{k+\ell,k} & \mathcal{K}_{k+\ell,k+1} & \cdots & \mathcal{K}_{k+\ell,k+\ell} \end{bmatrix}$$

and

$$\mathcal{D}_{k,\ell} = egin{bmatrix} \mathcal{I} & & & \ & \mathcal{I} & & \ & & \ddots & \ & & & \mathcal{I} \end{bmatrix},$$

where the \mathcal{I} 's on the diagonal are identity operators on the corresponding subspaces \mathbb{F}_k , $\mathbb{W}_{k+1}, \ldots, \mathbb{W}_{k+\ell}$. Note

$$\mathcal{D}_{k,\ell} = \mathcal{I} - \mathcal{A}_{k,\ell} + \mathcal{U}_{k,\ell} + \mathcal{L}_{k,\ell}$$

and it is in fact the operator $\mathcal{I} - \mathcal{K}_k$ on X. Now, equation (4.125) becomes

$$(\mathcal{D}_{k,\ell} - \mathcal{U}_{k,\ell} - \mathcal{L}_{k,\ell})\mathbf{u}_{k,\ell} = \mathbf{f}_{k,\ell}.$$
(4.126)

(4.126) suggests following two iterative schemes in terms of operator forms:

• Jacobi type iteration:

$$\mathcal{D}_{k,\ell}\mathbf{u}_{k,\ell}^{(\mu+1)} = (\mathcal{U}_{k,\ell} + \mathcal{L}_{k,\ell})\mathbf{u}_{k,\ell}^{(\mu)} + \mathbf{f}_{k,\ell}, \quad \mu \in \mathbb{N}_0$$
(4.127)

with any initial approximation $\mathbf{u}_{k,\ell}^{(0)}$.

• Gauss-Seidel type iteration:

$$(\mathcal{D}_{k,\ell} - \mathcal{U}_{k,\ell})\mathbf{u}_{k,\ell}^{(\mu)} = \mathcal{L}_{k,\ell}\mathbf{u}_{k,\ell}^{(\mu)} + \mathbf{f}_{k,\ell}, \quad \mu \in \mathbb{N}_0$$
(4.128)

with any initial approximation $\mathbf{u}_{k,\ell}^{(0)}$.

Following theorem says that above iterative schemes are convergent for appropriately chosen k.

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Theorem 4.28 Assume that the operator $\mathcal{I} - \mathcal{K}$ is bijective on a Banach space \mathbb{X} with \mathcal{K} being compact on \mathbb{X} , and the sequence of projections $\mathcal{P}_n, n \in \mathbb{N}_0$ satisfies the nestedness property (4.120). Then, for each $\ell \geq 1$ and sufficiently large k, the iteration schemes (4.127) and (4.128) for solving the equation (4.116) are convergent.

Proof: This theorem is a generalization of Theorem 3.1 in [FMX]. We consider Banach space in this thesis instead of Hilbert space in [FMX], hence the projections are different. Recall the definition of operator \mathcal{P}_n , $n \in \mathbb{N}_0$ (3.52) and the property of pointwise convergence of \mathcal{P}_n (3.53). By (4.119), we have for i > j,

$$\lim_{n \to \infty} \left\| (\mathcal{P}_{n+i} - \mathcal{P}_{n+j}) \mathcal{K} \right\| = \lim_{n \to \infty} \left\| \mathcal{K} (\mathcal{P}_{n+i} - \mathcal{P}_{n+j}) \right\| = 0.$$

Note that, for all $u \in \mathbb{X}$,

$$\mathcal{U}_{k,\ell}u = \mathcal{U}_{k,\ell}(\mathcal{P}_{k+\ell} - \mathcal{P}_k)u,$$

and

$$\mathcal{U}_{k,\ell} u = \mathcal{P}_k \mathcal{K}(\mathcal{P}_{k+\ell} - \mathcal{P}_k)u + \sum_{i=1}^{\ell-1} \mathcal{Q}_{k+i} \mathcal{K}(\mathcal{P}_{k+\ell} - \mathcal{P}_{k+i})u.$$

Hence, for each fixed $\ell \geq 1$,

$$\|\mathcal{U}_{k,\ell}\| \le C \sum_{j=0}^{\ell-1} \|\mathcal{K}(\mathcal{P}_{k+\ell} - \mathcal{P}_{k+j})\| \to 0 \quad \text{as} \quad k \to \infty,$$
(4.129)

where we have used the fact that $\|\mathcal{P}_j\| \leq C, j \geq 0$ and $\|\mathcal{Q}_i\| \leq C, i \geq 1$ (due to the pointwise convergence of $\mathcal{P}_n, n \in \mathbb{N}_0$ and the principle of uniformly boundedness). Similarly, we have

$$\mathcal{L}_{k,\ell} u = (\mathcal{P}_{k+\ell} - \mathcal{P}_k)\mathcal{K}\mathcal{P}_k u + \sum_{i=0}^{\ell-1} (\mathcal{P}_{k+\ell} - \mathcal{P}_{k+i})\mathcal{K}\mathcal{Q}_{k+1+i} u,$$

and thus

$$\|\mathcal{L}_{k,\ell}\| \le C \sum_{i=0}^{\ell-1} \|(\mathcal{P}_{k+\ell} - \mathcal{P}_{k+i})\mathcal{K}\| \to 0 \quad \text{as} \quad k \to \infty.$$

Moreover, since $\mathcal{U}_{k,\ell} + \mathcal{L}_{k,\ell} = \mathcal{K}_{k+\ell} - \mathcal{K}_k$,

$$\|\mathcal{U}_{k,\ell} + \mathcal{L}_{k,\ell}\| = \|(\mathcal{P}_{k+\ell} - \mathcal{P}_k)\mathcal{K}\| \to 0 \quad \text{as} \quad k \to \infty.$$
(4.130)

On the other hand,

$$\mathcal{D}_{k,\ell}^{-1} = (\mathcal{I} - \mathcal{K}_k)^{-1} \quad \text{and} \quad (\mathcal{D}_{k,\ell} - \mathcal{U}_{k,\ell})^{-1} = (\mathcal{I} - (\mathcal{I} - \mathcal{K}_k)^{-1}\mathcal{U}_{k,\ell})^{-1} (\mathcal{I} - \mathcal{K}_k)^{-1}.$$

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Therefore,

$$\|\mathcal{D}_{k,\ell}^{-1}\| = \|(\mathcal{I} - \mathcal{K}_k)^{-1}\| \quad \text{and} \quad \|(\mathcal{D}_{k,\ell} - \mathcal{U}_{k,\ell})^{-1}\| \le \frac{\|(\mathcal{I} - \mathcal{K}_k)^{-1}\|}{1 - \|(\mathcal{I} - \mathcal{K}_k)^{-1}\|\|\mathcal{U}_{k,\ell}\|}.$$
 (4.131)

Since $\|(\mathcal{I} - \mathcal{K}_k)^{-1}\|$ is uniformly bounded for large enough k, in view of (4.129), these two inverses are both bounded uniformly for large enough k. Thus, for each fixed $\ell \geq 1$, the iteration operators for schemes (4.127) and (4.128) are both convergent to 0 in operator norm:

$$\|\mathcal{D}_{k,\ell}^{-1}(\mathcal{L}_{k,\ell}+\mathcal{U}_{k,\ell})\| \to 0 \text{ and } \|(\mathcal{D}_{k,\ell}-\mathcal{U}_{k,\ell})^{-1}\mathcal{L}_{k,\ell}\| \to 0 \text{ as } k \to \infty.$$

Therefore they can be less than 1 if k is large enough, which in turn yields the convergence of the iteration schemes (4.127) and (4.128). \Box

We now describe above iterative methods in terms of their matrix representations. The matrix representation of (4.116) is

$$(\mathbf{E}_n - \mathbf{K}_n)\mathbf{u}_n = \mathbf{f}_n. \tag{4.132}$$

Let $\mathbf{A}_n = \mathbf{E}_n - \mathbf{K}_n$. We begin with an initial level k: $1 \leq k < n$ (we have seen, for the convergence of the iteration, k needs to not too small), and let $\ell = n - k$ be the number of additional levels from level k to the final level n. Then \mathbf{A}_n is partitioned into $(\ell+1) \times (\ell+1)$ -block matrix:

$$\mathbf{A}_n = [\mathbf{A}_{ij} : i, j = k, \dots, k + \ell].$$

Introduce following notations for the convenience of discussion:

$$\mathbf{U}_{k,\ell} = \begin{bmatrix} \mathbf{E}_k - \mathbf{K}_k & & \\ & \mathbf{I} & \\ & & \ddots & \\ & & & \mathbf{I} \end{bmatrix},$$
$$\mathbf{U}_{k,\ell} = \begin{bmatrix} O & \mathbf{E}_{k,k+1} - \mathbf{K}_{k,k+1} & \cdots & \mathbf{E}_{k,k+\ell} - \mathbf{K}_{k,k+\ell} \\ O & \cdots & \mathbf{E}_{k+1,k+\ell} - \mathbf{K}_{k+1,k+\ell} \\ & \ddots & \vdots \\ & & O \end{bmatrix},$$

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and

$$\mathbf{L}_{k,\ell} = \begin{bmatrix} O \\ \mathbf{K}_{k+1,k} & \mathbf{K}_{k+1,k+1} \\ \vdots & \vdots & \ddots \\ \mathbf{K}_{k+\ell,k} & \mathbf{K}_{k+\ell,k+1} & \cdots & \mathbf{K}_{k+\ell,k+\ell} \end{bmatrix}$$

Accordingly, (4.132) is written as

$$(\mathbf{D}_{k,\ell} + \mathbf{U}_{k,\ell} - \mathbf{L}_{k,\ell})\mathbf{u}_{k,\ell} = \mathbf{f}_{k,\ell}.$$
(4.133)

The two multilevel iteration schemes (4.127) and (4.128) in terms of matrix representation now have forms:

• Jacobi type iteration:

$$\mathbf{D}_{k,\ell}\mathbf{u}_{k,\ell}^{(\mu+1)} = (-\mathbf{U}_{k,\ell} + \mathbf{L}_{k,\ell})\mathbf{u}_{k,\ell}^{(\mu)} + \mathbf{f}_{k,\ell}, \quad \mu \in \mathbb{N}_0$$
(4.134)

with any initial guess $\mathbf{u}_{k,\ell}^{(0)}$. The iteration matrix for Jacobi type scheme is

$$\mathbf{M}_{k,\ell}^{\mathrm{J}} = \mathbf{D}_{k,\ell}^{-1}(-\mathbf{U}_{k,\ell} + \mathbf{L}_{k,\ell})$$

• Gauss-Seidel type iteration:

$$(\mathbf{D}_{k,\ell} + \mathbf{U}_{k,\ell})\mathbf{u}_{k,\ell}^{(\mu+1)} = \mathbf{L}_{k,\ell}\mathbf{u}_{k,\ell}^{(\mu)} + \mathbf{f}_{k,\ell}, \quad \mu \in \mathbb{N}_0$$
(4.135)

with any initial guess $\mathbf{u}_{k,\ell}^{(0)}$. The iteration matrix for Gauss-Seidel type scheme is:

$$\mathbf{M}_{k,\ell}^{\mathrm{GS}} = (\mathbf{D}_{k,\ell} + \mathbf{U}_{k,\ell})^{-1} \mathbf{L}_{k,\ell}$$

Both (4.134) and (4.135) can be solved easily by "backward substitutions" for all blocks except the first block of $\mathbf{u}_{k,\ell}^{(\mu+1)}$. For the first block of $\mathbf{u}_{k,\ell}^{(\mu+1)}$, we need to find $(\mathbf{E}_k - \mathbf{K}_k)^{-1}$. In comparison with the Jacobi type iterate, Gauss-Seidel type scheme uses previously computed components within each iteration to compute the next component by backward substitution, while the Jacobi scheme doesn't use the new computed components until an iteration is completed. Both methods require inverting the submatrix $\mathbf{E}_k - \mathbf{K}_k$. Hence, we need to choose k in order that the inverse of $\mathbf{E}_k - \mathbf{K}_k$ exists and is relatively easy to find, i.e, k

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		Multilevel iterate			Gaussian elimination			
	(k,ℓ)	μ	$\ u_{k,\ell}^{(\mu)} - u\ _{\infty}$	order	time(sec.)	$\ u_{k+\ell}^* - u\ _{\infty}$	order	time(sec.)
ſ	(3,0)		7.499082e-3		0.16	7.499082e-3		0.16
	(3,1)	4	1.845814e-3	2.0230	0.81	1.849162e-3	2.0198	10
	(3,2)	8	4.650403e-4	1.9888	8.74	4.650610e-4	1.9914	632

Table 4.14: Comparison of uncompressed solutions using Gaussian elimination and multilevel iteration methods.

is not too large (for example k = 2 or 3). Conveniently, a good initial guess would be $\mathbf{u}_{k,\ell}^{(0)} := [(\mathbf{E}_k - \mathbf{K}_k)^{-1} \mathbf{f}_k, 0, \dots, 0]^T$. Another issue is that we need to choose a proper k such that the iteration scheme is convergent. Namely, we need the spectral radius of $\mathbf{M}_{k,\ell}^{\text{GS}}$ (or $\mathbf{M}_{k,\ell}^{\text{J}}$) is less than one. This can be achieved by requiring their matrix norms are less than one. Our experiments shows when the domain of integration is unit triangle Ω , then k must be ≥ 3 , and for other domains of integration (for example, a general triangle, a polygon), $k \geq 2$.

Example 4.29 In this example, we test if the Gauss-Seidel type multilevel iteration method gives good solution. The results are compared with the results from stand Gaussian elimination. All the entries are computed analytically and no compression is assumed. The results are tabulated in Table 4.14. We can see that both methods give very close solutions. Where, μ is the number of iterations. The significant reduction of the computing time is very impressive.

Chapter 5

Fast Wavelet Collocation Methods on Polygonal Domains

5.1 On General Triangles

Let \triangle denote a planar triangle in the *xy*-plane, and let vertices of \triangle be denoted by $\{v_j, j \in \mathbb{Z}_3\}$, where, $v_j := (x_j, y_j)$. As before, Ω denotes the unit triangle, i.e.,

$$\Omega = \{(s,t)|s,t \ge 0, s+t \le 1\}.$$
(5.1)

The vertices of the unit triangle Ω are

$$q_0 = (0,0), \quad q_1 = (1,0), \quad q_2 = (0,1).$$
 (5.2)

Define an affine mapping

$$T: \Omega \xrightarrow[onto]{honto} \Delta$$

by

$$(x,y) = T(s,t) := (1-s-t)v_0 + sv_1 + tv_2,$$
(5.3)

namely

$$\begin{cases} x = (1-s-t)x_0 + sx_1 + tx_2 \\ y = (1-s-t)y_0 + sy_1 + ty_2 \end{cases}$$
(5.4)

Note $T(q_i) = v_i$, $i \in \mathbb{Z}_3$. The inverse of this mapping is given by solving (5.3) for (s, t), or equivalently by solving

$$\begin{cases} t(x_2 - x_0) + s(x_1 - x_0) = x - x_0 \\ t(y_2 - y_0) + s(y_1 - y_0) = y - y_0 \end{cases}$$
(5.5)

for (s, t). This system is nonsingular since the vertices $\{v_i, i \in \mathbb{Z}_3\}$ are assumed not to be collinear. Denote the inverse mapping of T by T^{-1} , then

$$(s,t) = T^{-1}(x,y). (5.6)$$

 T^{-1} is an affine mapping too. Let

$$A = \det \begin{pmatrix} x_1 - x_0 & x_2 - x_0 \\ y_1 - y_0 & y_2 - y_0 \end{pmatrix}.$$
 (5.7)

 T^{-1} has an explicit expression, namely,

$$\begin{cases} s = \frac{1}{A} [x(y_2 - y_0) + y(x_0 - x_2) + (x_2y_0 - x_0y_2)] \\ t = \frac{1}{A} [x(y_0 - y_1) + y(x_1 - x_0) + (x_0y_1 - x_1y_0)]. \end{cases}$$
(5.8)

Observe that the Jacobian of T is $J_T = |A|$ and the Jacobian of T^{-1} is $J_{T^{-1}} = \frac{1}{|A|} = \frac{1}{J_T}$. Also $J_T = 2\operatorname{Area}(\Delta)$.

Let $\mathbb{X} := L^{\infty}(\Delta)$ and $\mathbb{V} := C(\Delta)$. We now consider the integral equation

$$u(x,y) - \int_E K(x,y,\xi,\eta)u(\xi,\eta)\mathrm{d}\xi\mathrm{d}\eta = f(x,y),\tag{5.9}$$

where, $K(x, y, \xi, \eta)$ is a weakly singular kernel. E here is the general triangle \triangle . Therefore the operator $\mathcal{K} : \mathbb{X} \to \mathbb{V}$ defined by

$$(\mathcal{K}u)(x,y) := \int_{\Delta} K(x,y,\xi,\eta) u(\xi,\eta) \mathrm{d}\xi \mathrm{d}\eta, \quad (x,y) \in \Delta$$
(5.10)

is compact in X. u is the unknown in X. Use the transformation (x, y) = T(s, t), (5.9) changes into

$$u(s,t) - J_T \int_{\Omega} K(T(s,t), T(s',t')) u(s',t') ds' dt' = f(T(s,t)).$$
(5.11)

Note that (5.11) is just an integral equation defined on the unit triangle, thus we can use the method described in Chapter 3 and Chapter 4 to solve (5.11). Assume the wavelet

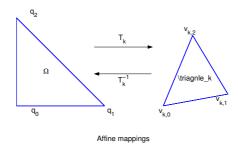


Figure 5.1: The affine mapping T_k

collocation solution to (5.11) on Ω is $u_n(s,t) = \sum_{(i,j) \in \mathbb{U}_n} c_{ij} w_{ij}(s,t)$, then the collocation solution to (5.9) on \triangle can be obtained by

$$u_n(x,y) = \sum_{(i,j)\in\mathbb{U}_n} c_{ij} w_{ij}(T^{-1}(x,y)).$$
(5.12)

5.2 On Polygons

Given a polygonal domain E in the plane \mathbb{R}^2 , and let $\triangle_{\mathbf{p}} := \{ \triangle_0, \dots, \triangle_{p-1} \}$ denote a triangulation of E. We assume triangles \triangle_j and \triangle_k $(j, k \in \mathbb{Z}_p, j \neq k)$ can intersect only at vertices or along their common edges. Define affine mappings (see Fig. 5.1)

$$T_k: \ \Omega \mapsto \triangle_k \quad k \in \mathbb{Z}_p.$$
(5.13)

Their expressions can be readily obtained from (5.3) or (5.4). Denote the three vertices of Δ_k by $\{v_{k,i}, i \in \mathbb{Z}_3\}$ with

$$v_{k,i} = T_k(q_i), \quad i \in \mathbb{Z}_3, \tag{5.14}$$

where $v_{k,i} = (x_{k,i}, y_{k,i})$.

We now develop the wavelet collocation method for the integral equation (5.9), but in this context, E is a polygonal domain and $\mathbb{X} = L^{\infty}(E)$, $\mathbb{V} = C(E)$. Note $E = \bigcup_{k=0}^{p-1} \Delta_k$. Observe that

$$\Phi(T_k^{-1}\Delta_k) = T_k^{-1}\Delta_k, \quad k \in \mathbb{Z}_p,$$
(5.15)

where $\Phi = \{\phi_e, e \in \mathbb{Z}_4\}$ is defined as in (3.1). Hence we have

$$T_k \circ \Phi \circ T_k^{-1} \triangle_k = \triangle_k. \tag{5.16}$$

For convenience, we define

$$\phi_{k,e} := T_k \circ \phi_e \circ T_k^{-1}, \quad k \in \mathbb{Z}_p, \ e \in \mathbb{Z}_4, \tag{5.17}$$

and

$$\Phi_k := T_k \circ \Phi \circ T_k^{-1}, \quad k \in \mathbb{Z}_p.$$
(5.18)

We see from (5.16), the set Δ_k , $k \in \mathbb{Z}_p$ is the invariant set associated with the family of mappings Φ_k . Define

$$\hat{\Phi} := \{ \Phi_k : \ k \in \mathbb{Z}_p \} \tag{5.19}$$

Then we readily obtain by (5.16) that

$$\hat{\Phi}(E) = E. \tag{5.20}$$

Namely, E is an invariant set with respect to the family of contractive mappings $\hat{\Phi}$. We now determine the initial interpolating points for $\Delta_k \ k \in \mathbb{Z}_p$. Let

$$G_{k,0} = \{ t_{0j}^{(k)} : j \in \mathbb{Z}_3, \ k \in \mathbb{Z}_p \}$$
(5.21)

with $t_{0j}^{(k)} = T_k t_{0j}, \ j \in \mathbb{Z}_3, k \in \mathbb{Z}_p$, and $t_{0j} \in \Omega$ are defined in Chapter 3. Note $G_{k,0}$ is refinable relative to the mappings $\Phi_k, k \in \mathbb{Z}_p$, i.e., $G_{k,0}$ satisfies

$$G_{k,0} \subset \Phi_k(G_{k,0}). \tag{5.22}$$

Define $G_0 := \bigcup_{k \in \mathbb{Z}_p} G_{k,0}$. Therefore, we obtain

$$G_0 \subset \hat{\Phi}(G_0). \tag{5.23}$$

i.e., G_0 is refinable relatively with respect to the family of contractive mappings $\hat{\Phi}$. Now, we construct scaling functions and initial wavelets $w_{ij}^{(k)}$, $i \in \mathbb{Z}_2$ which are supported on Δ_k by

$$w_{ij}^{(k)}(x,y) = \begin{cases} w_{ij} \circ T_k^{-1}(x,y) & (x,y) \in \Delta_k \\ 0 & (x,y) \in E \setminus \Delta_k, \end{cases}$$
(5.24)

where, when $i = 0, j \in \mathbb{Z}_3$ and, when $i = 1, j \in \mathbb{Z}_9$. Moreover, define

$$\mathbb{W}_0 = \operatorname{span}\{w_{0j}^{(k)}: \ j \in \mathbb{Z}_3, \ k \in \mathbb{Z}_p\}$$

$$(5.25)$$

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and

$$\mathbb{W}_{1} = \operatorname{span}\{w_{1j}^{(k)}: j \in \mathbb{Z}_{9}, k \in \mathbb{Z}_{p}\}.$$
(5.26)

We now construct initial collocation functionals. Let

$$\ell_{0j}^{(k)} := \delta_{t_{0j}^{(k)}}, \text{ with } t_{0j}^{(k)} = T_k(t_{0j}), \quad j \in \mathbb{Z}_3, k \in \mathbb{Z}_p$$
(5.27)

and

$$\ell_{1j}^{(k)} = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} \delta_{t_{1l}^{(k)}}, \text{ with } t_{1l}^{(k)} = T_k(t_{1l}), \quad j \in \mathbb{Z}_9, k \in \mathbb{Z}_p,$$
(5.28)

where, $t_{1l} \in \Omega$, $l \in \mathbb{Z}_{12}$ are defined in (3.30) of Chapter 3, and c'_{jl} , $j \in \mathbb{Z}_9$, $l \in \mathbb{Z}_{12}$ are elements of **C'** in (3.33). In order to generate the multiscale wavelets for the space \mathbb{W}_i , $i \geq 2$, we define linear operators $\mathcal{T}_{k,e} : \mathbb{X}_k \to \mathbb{X}_k$, $e \in \mathbb{Z}_4$, $k \in \mathbb{Z}_p$ by

$$(T_{k,e}x)(t) := x(\phi_{k,e}^{-1}(t))\chi_{\phi_{k,e}(\triangle_k)}(t),$$
(5.29)

where, $\mathbb{X}_k := L^{\infty}(\Delta_k), \ k \in \mathbb{Z}_p$. For $\mathbf{e} := (e_0, \ldots, e_{n-1}) \in \mathbb{Z}_4^n$ we introduce a composition operator $\mathcal{T}_{k,\mathbf{e}}$ by

$$\mathcal{T}_{k,\mathbf{e}} := \mathcal{T}_{k,e_0} \circ \ldots \circ \mathcal{T}_{k,e_{n-1}}.$$
(5.30)

Now, for i = 2, 3, ..., n, let

$$w_{ij}^{(k)}(x,y) = \begin{cases} \mathcal{T}_{k,\mathbf{e}} w_{1l}^{(k)}(x,y) & (x,y) \in \Delta_k \\ 0 & (x,y) \in E \setminus \Delta_k, \end{cases}$$
(5.31)

where, $j = 9\mu(\mathbf{e}) + l$, $\mathbf{e} \in \mathbb{Z}_4^{i-1}$, $l \in \mathbb{Z}_9$. It follows that

$$\mathbb{W}_i := \operatorname{span}\{w_{ij}^{(k)} : j \in \mathbb{Z}_{w(i)}, \ k \in \mathbb{Z}_p\}.$$
(5.32)

Observe that the support of $w_{ij}^{(k)}$ is contained in $S_{ij}^{(k)} := \phi_{k,\mathbf{e}}(\Delta_k), \ j \in \mathbb{Z}_{w(i)}$. To generate a multiscale collocation functionals, we introduce for any $e \in \mathbb{Z}_4$ a linear operator $\mathcal{L}_{k,e} : \mathbb{X}_k^* \to \mathbb{X}_k^*$ defined by the equation

$$\langle \mathcal{L}_{k,e}\ell, v \rangle = \langle \ell, v \circ \phi_{k,e} \rangle, \quad v \in \mathbb{X}_k, \ell \in \mathbb{X}_k^*.$$
 (5.33)

Moreover, for $\mathbf{e} := (e_0, \ldots, e_{n-1}) \in \mathbb{Z}_4^n$, we define the composite operator

$$\mathcal{L}_{k,\mathbf{e}} := \mathcal{L}_{k,e_0} \circ \ldots \circ \mathcal{L}_{k,e_{n-1}}.$$
(5.34)

Consequently, for any $\mathbf{e}, \mathbf{e}' \in \mathbb{Z}_4^i$, $w \in \mathbb{X}_k$ and $\ell \in \mathbb{X}_k^*$, we have that

$$\langle \mathcal{L}_{k,\mathbf{e}}l, \mathcal{T}_{k,\mathbf{e}'}w \rangle = \langle l, w \rangle \delta_{\mathbf{e}\mathbf{e}'}.$$
 (5.35)

Now, for i > 1, $j = 9\mu(\mathbf{e}) + l$, $\mathbf{e} \in \mathbb{Z}_4^{i-1}$, $l \in \mathbb{Z}_9$ define

$$\ell_{ij}^{(k)} := \mathcal{L}_{k,\mathbf{e}} \ell_{1l}^{(k)}, \tag{5.36}$$

and observe that

$$\langle \ell_{ij}^{(k)}, v \rangle = \langle \ell_{1l}^{(k)}, v \circ \phi_{k,\mathbf{e}} \rangle = \sum_{s \in \mathbb{Z}_{12}} c'_{ls} v(\phi_{k,\mathbf{e}}(t_{1s}^{(k)})).$$
 (5.37)

Note that the "support" of $\ell_{ij}^{(k)}$ is also contained in $S_{ij}^{(k)}$. We now discover some close ties between $\ell_{ij}^{(k)}$, $w_{ij}^{(k)}$, $k \in \mathbb{Z}_p$, $(i, j) \in \mathbb{U}$ defined on Δ_k and ℓ_{ij} , w_{ij} , $(i, j) \in \mathbb{U}$ defined on Ω . First observe that from (5.17), for any $\mathbf{e} \in \mathbb{Z}_{\mu}^n$ and $k \in \mathbb{Z}_p$,

$$\phi_{k,\mathbf{e}} = T_k \circ \phi_{\mathbf{e}} \circ T_k^{-1} \tag{5.38}$$

and

$$T_k^{-1} \circ \phi_{k,e} \circ T_k = \phi_e, \qquad T_k^{-1} \circ \phi_{k,\mathbf{e}} \circ T_k = \phi_{\mathbf{e}}.$$
(5.39)

Following two lemmas are important for the discussion of properties of $w_{ij}^{(k)}$ and $\ell_{ij}^{(k)}$.

Lemma 5.1

$$\langle \ell_{ij}^{(k)}, v \rangle = \langle \ell_{ij}, v \circ T_k \rangle, \tag{5.40}$$

where, $v \in L_{\infty}(\Delta_k)$, $k \in \mathbb{Z}_p$, $(i, j) \in \mathbb{U}$.

Proof: For $j = \mu(\mathbf{e})r + l$, $l \in \mathbb{Z}_r$, with r = 9,

$$\langle \ell_{ij}^{(k)}, v \rangle = \langle \mathcal{L}_{k,\mathbf{e}} \ell_{1l}^{(k)}, v \rangle$$

$$= \langle \ell_{1l}^{(k)}, v \circ \phi_{k,\mathbf{e}} \rangle = \sum_{s \in \mathbb{Z}_{12}} c_{ls}' v \circ \phi_{k,\mathbf{e}}(t_{1s}^{(k)})$$

$$= \sum_{s \in \mathbb{Z}_{12}} c_{ls}' v \circ \phi_{k,\mathbf{e}}(T_k t_{1s}) = \sum_{s \in \mathbb{Z}_{12}} c_{ls}' v \circ T_k \circ \phi_{\mathbf{e}} \circ T_k^{-1} \circ T_k(t_{1s})$$

$$= \sum_{s \in \mathbb{Z}_{12}} c_{ls}' (v \circ T_k) \circ \phi_{\mathbf{e}}(t_{1s}) = \langle \ell_{ij}, v \circ T_k \rangle.$$

$$(5.41)$$

Lemma 5.2

$$w_{ij}^{(k)}(x,y) = w_{ij} \circ T_k^{-1}(x,y).$$
(5.42)

Proof: For $j = \mu(\mathbf{e})r + l$, $l \in \mathbb{Z}_r$, with r = 9,

$$w_{ij}^{(k)}(x,y) = \mathcal{T}_{k,\mathbf{e}} w_{1l}^{(k)}(x,y) = w_{1l}^{(k)} \circ \phi_{k,\mathbf{e}}^{-1}(x,y) \chi_{\phi_{k,\mathbf{e}}(\Delta_{k})}(x,y) = (w_{1l}^{(k)} \circ T_{k} \circ \phi_{\mathbf{e}}^{-1}) \circ T_{k}^{-1}(x,y) \chi_{T_{k} \circ \phi_{\mathbf{e}} \circ T_{k}^{-1}(\Delta_{k})}(x,y) = (w_{1l} \circ \phi_{\mathbf{e}}^{-1}) \circ T_{k}^{-1}(x,y) \chi_{T_{k} \circ \phi_{\mathbf{e}} \circ T_{k}^{-1}(\Delta_{k})}(x,y) = (w_{1l} \circ \phi_{\mathbf{e}}^{-1} \chi_{\phi_{\mathbf{e}}(\Omega)}) \circ T_{k}^{-1}(x,y) = w_{ij} \circ T_{k}^{-1}(x,y),$$
(5.43)

where we have used $\chi_{T_k\phi_{\mathbf{e}}T_k^{-1}(\triangle_k)}(x,y) = \chi_{\phi_{\mathbf{e}}(\Omega)}(s,t)$ with $(x,y) = T_k(s,t), (x,y) \in \triangle_k, (s,t) \in \Omega$.

Remark: Lemma 5.1 and Lemma 5.2 provide alternative ways to construct ℓ_{ij} and w_{ij} for $(i, j) \in \mathbb{U}_n, i \geq 2$ as compared to (5.31) and (5.36).

5.3 Computation of Matrix Elements

We now consider in detail the discretization of the integral equation on a polygonal domain E. Consider

$$u(x,y) - \int_E K(x,y,\xi,\eta) \mathrm{d}\xi \mathrm{d}\eta = f(x,y), \qquad (5.44)$$

where E is the polygonal domain, and E has a triangulation $\triangle_{\mathbf{p}}$ such that $E = \bigcup_{k \in \mathbb{Z}_p} \triangle_k$. Let T_k be the mapping associated with each triangle \triangle_k , such that

$$(x,y) = T_k(s,t), \quad (x,y) \in \Delta_k, \ (s,t) \in \Omega.$$
(5.45)

Let u_n be the approximation solution to u in \mathbb{F}_n with

$$u_n(x,y) = \sum_{k \in \mathbb{Z}_p} \sum_{i,j} c_{ij}^{(k)} w_{ij}^{(k)}(x,y), \quad i \in \mathbb{Z}_{n+1}, \ j \in \mathbb{Z}_{w(i)}.$$
(5.46)

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By using (5.45) we then have

$$u_n(s,t) = \sum_{k \in \mathbb{Z}_p} \sum_{i,j} c_{ij}^{(k)} w_{ij}(s,t), \quad i \in \mathbb{Z}_{n+1}, \ j \in \mathbb{Z}_{w(i)}, \quad (s,t) \in \Omega$$
(5.47)

Now use u_n replacing u in (5.44), we have

$$\sum_{k' \in \mathbb{Z}_p} \sum_{(i,j) \in \mathbb{U}_n} c_{ij}^{(k')} w_{ij}(s,t) - \sum_{k' \in \mathbb{Z}_p} J_{T_{k'}} \sum_{(i,j) \in \mathbb{U}_n} c_{ij}^{(k')} \int_{\Omega} K(T_k(s,t), T_{k'}(s',t')) w_{ij}(s',t') \mathrm{d}s' \mathrm{d}t'$$

= $f \circ T_k(s,t),$ (5.48)

where, $k \in \mathbb{Z}_p$. Now apply $\ell_{i'j'}^{(k)}, k \in \mathbb{Z}_p, (i', j') \in \mathbb{U}_n$ to both sides of equation (5.48), we have

$$\left\langle \ell_{i'j'}^{(k)}, \sum_{k' \in \mathbb{Z}_p} \sum_{(i,j) \in \mathbb{U}_n} c_{ij}^{(k')} w_{ij} \right\rangle - \sum_{k' \in \mathbb{Z}_p} J_{T_{k'}} \sum_{(i,j) \in \mathbb{U}_n} c_{ij}^{(k')} \int_{\Omega} \left\langle \ell_{i'j'}^{(k)}, K\left(T_k(\cdot, \cdot), T_{k'}(s', t')\right) \right\rangle w_{ij}(s', t') \mathrm{d}s' \mathrm{d}t' = \left\langle \ell_{i'j'}^{(k)}, f \circ T_k(s, t) \right\rangle$$

$$(5.49)$$

After discretization, we obtain a system of linear equations

$$(\mathbf{E}_n - \mathbf{K}_n)\mathbf{u}_n = \mathbf{f}_n,\tag{5.50}$$

where

$$\mathbf{E}_{n} = [E_{k'i'j',kij}] = [\langle \ell_{i'j'}^{(k')}, w_{ij}^{(k)} \rangle : k, k' \in \mathbb{Z}_{p}, (i,j), (i',j') \in \mathbb{U}_{n}],$$

$$\mathbf{K}_{n} = [K_{k'i'j',kij} : k, k' \in \mathbb{Z}_{p}, (i, j), (i', j') \in \mathbb{U}_{n}] \\ = \left[J_{T_{k}} \int_{\Omega} \langle \ell_{i'j'}^{(k')} K(T_{k'}(\cdot, \cdot), T_{k}(s', t')) \rangle w_{ij}(s', t') \mathrm{d}s' \mathrm{d}t' : k, k' \in \mathbb{Z}_{p}, (i, j), (i', j') \in \mathbb{U}_{n} \right],$$

and

$$\mathbf{f}_n = [f_{k'ij}] = \left[\langle \ell_{i'j'}^{(k')}, fT_{k'}(s,t) \rangle : k' \in \mathbb{Z}_p, (i',j') \in \mathbb{U}_n \right].$$

5.4 Properties of the Wavelet Basis functions and the Collocation Functionals Over Polygons

Next several lemmas tells us the vanishing moments and orthogonality of the initial collocation functionals $\ell_{ij}^{(k)}$ and wavelets $w_{ij}^{(k)}$, where $(i, j) \in \mathbb{U}_2$, $k \in \mathbb{Z}_p$. Yi Wang Chapter 11. Fast wavelet collocation method on a polygonal domain 96

Lemma 5.3

$$\langle \ell_{0j}^{(k)}, w_{0j'}^{(k')} \rangle = \delta_{kk'} \delta_{jj'}, \quad j, j' \in \mathbb{Z}_3, \ k \in \mathbb{Z}_p.$$

$$(5.51)$$

Proof: Clearly, when $k \neq k'$, $\langle \ell_{0j}^k, w_{0j'}^{(k')} \rangle = 0$ by the definition of $\ell_{0j}^{(k)}$ and $w_{0j'}^{(k')}$. When k = k', $\langle \ell_{0j}^{(k)}, w_{0j'}^{(k')} \rangle = w_{0j'}^{(k)}(t_{0j}^{(k)}) = w_{0j'} \circ T_k^{-1} \circ T_k(t_{0j}) = w_{0j'}(t_{0j}) = \delta_{j'j}$.

Lemma 5.4

$$\langle \ell_{1j}^{(k)}, w_{1j'}^{(k')} \rangle = \delta_{kk'} \delta_{jj'}, \quad j, j' \in \mathbb{Z}_9; \quad k, k' \in \mathbb{Z}_p.$$

$$(5.52)$$

Proof: Clearly, when $k \neq k'$, $\langle \ell_{1j}^k, w_{1j'}^{(k')} \rangle = 0$ by the definition of $\ell_{1j}^{(k)}$ and $w_{1j'}^{(k')}$. When k = k', $\langle \ell_{1j}^{(k)}, w_{1j'}^{(k')} \rangle = \langle \ell_{1j}^{(k)}, w_{1j'} \circ T_k^{-1} \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} w_{1j'} \circ T_k^{-1} \circ (T_k t_{1l}) = \langle \ell_{1j}, w_{1j'} \rangle = \delta_{jj'}$.

Lemma 5.5

$$\langle w_{1j}^{(k)}, 1 \rangle = 0, \quad j \in \mathbb{Z}_9; \quad k \in \mathbb{Z}_p.$$

$$(5.53)$$

Proof: $\langle w_{1j}^{(k)}, 1 \rangle = \int_{\Delta_k} w_{1j}^{(k)} \cdot 1 dx dy = \int_{\Delta_k} w_{1j} \circ T_k^{-1}(x, y) dx dy = J_{T_{\Delta_k}} \int_{\Omega} w_{1j} \circ T_k^{-1} \circ T_k(s, t) ds dt = 0.$

Lemma 5.6

$$\langle w_{1j}^{(k)}, x \rangle = 0$$
 (5.54)

and

$$\langle w_{1j}^{(k)}, y \rangle = 0 \tag{5.55}$$

where, $j \in \mathbb{Z}_9$, $k \in \mathbb{Z}_p$.

Proof: $\langle w_{1j}^{(k)}, x \rangle = \int_{\Delta_k} w_{1j}^{(k)} x dx dy = \int_{\Delta_k} w_{1j} \circ T_k^{-1}(x, y) x dx dy = J_{T_k} \int_{\Omega} w_{1j} \circ T_k^{-1} \circ T_k(s, t) [(1 - s - t)x_{k,0} + sx_{k,1} + tx_{k,2}] ds dt = 0$ by the vanishing moment property of w_{1j} on Ω (Lemma 3.3).

Similarly, we can obtain $\langle w_{1j}^{(k)}, y \rangle = 0.$

Lemma 5.7

$$\langle \ell_{1j}^{(k)}, 1 \rangle = 0, \quad j \in \mathbb{Z}_9; \quad k \in \mathbb{Z}_p.$$

$$(5.56)$$

Proof: $\langle \ell_{1j}^{(k)}, 1 \rangle = \langle \sum_{l \in \mathbb{Z}_{12}} c'_{jl} \delta_{t_{1l}^{(k)}}, 1 \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} \langle T_k(t_{1l}), 1 \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} = 0$ by Lemma 3.3.

Lemma 5.8

$$\langle \ell_{1j}^{(k)}, x \rangle = 0, \tag{5.57}$$

and

$$\langle \ell_{1j}^{(k)}, y \rangle = 0, \tag{5.58}$$

where $j \in \mathbb{Z}_9$ and $k \in \mathbb{Z}_p$.

Proof: First note $\langle \ell_{1j}^{(k)}, x \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} \langle \delta_{T_k(t_{1l})}, x \rangle$. Now define $g(x, y) := x = (1 - s - t) x_{k,0} + s x_{k,1} + t x_{k,2} := h(s,t)$, where $(s,t) \in \Omega$ and $(x,y) \in \Delta_k$. Thus, we have

$$g(x,y) = h \circ T_k^{-1}(x,y).$$
(5.59)

Therefore,

 $\sum_{l \in \mathbb{Z}_{12}} c'_{jl} \langle \delta_{T_k(t_{1l})}, x \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} \langle \delta_{T_k(t_{1l})}, h \circ T_k^{-1} \rangle = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} h \circ T_k^{-1} \circ (T_k t_{1l}) = \sum_{l \in \mathbb{Z}_{12}} c'_{jl} h(t_{1l}) = \langle \ell_{1j}, h \rangle = 0$ by the vanishing moment of ℓ_{1j} (Lemma 3.3) since h(s, t) is linear in s, t. Similarly, we can prove $\langle \ell_{1j}^{(k)}, y \rangle = 0$.

Lemma 5.9 For matrix $\mathbf{E}_n := [E_{k'i'j',kij}]$, where $i', i \in \mathbb{Z}_{n+1}, j \in \mathbb{Z}_{w(i)}, j' \in \mathbb{Z}_{w(i')}, k, k' \in \mathbb{Z}_p$, if $k \neq k'$, $E_{k'i'j',kij} = 0$. For k = k', if i' > i, $E_{k,i'j',kij} = 0$ if i' = i, $E_{ki'j',kij} = \delta_{i'i}\delta_{jj'}$ if i' < i, in each submatrix $\mathbf{E}_{ki',ki}$ with shape parameter θ , only the elements on the main

diagonal line are possibly nonzero (cf. Lemma 4.3).

Proof: When $k \neq k'$, by the definition of the wavelet $w_{ij}^{(k)}$ and collocation functional $\ell_{i'j'}^{(k')}$, $E_{k'i'j',kij} = 0$. When k = k' the proof is parallel to the proof of Lemma 4.3.

Lemma 5.10 through Lemma 5.17 tells us important properties of the constructed wavelet basis functions and the collocation functionals on polygons. They are essentially the same properties as those of the wavelet basis functions and the collocation functionals on the unit triangle.

Lemma 5.10 The constructed wavelet basis functions $w_{ij}^{(k)}$ over the polygon E are locally supported and their supports are shrinking as level i increases.

Proof: for every i > 1, the support of $w_{ij}^{(k)}$ is contained in $S_{ij}^{(k)} = \phi_{k,\mathbf{e}}(\triangle_k) = (\triangle_k)_{i-1,\mu(\mathbf{e})}$, where $j = \mu(\mathbf{e})r + l$, $l \in \mathbb{Z}_r$, with r = 9.

Lemma 5.11 For any $i, i' \in \mathbb{N}_0$,

$$\langle \ell_{i'j'}^{(k')}, w_{ij}^{(k)} \rangle = \delta_{k'k} \delta_{ii'} \delta_{jj'}, \quad (i,j), (i',j') \in \mathbb{U}, \ i \le i', \ k, k' \in \mathbb{Z}_p,$$
(5.60)

$$\sum_{j \in \mathbb{Z}_{w(i)}} |\langle \ell_{i'j'}^{(k)}, w_{ij}^{(k)} \rangle| \le \frac{81}{28}, \quad (i,j), (i',j') \in \mathbb{U}, i > i',$$
(5.61)

and if $\gamma := \frac{81}{28}$, then

$$\gamma < \mu^{s/2} - 1$$
, with $s = 2$, (5.62)

where, s is the order of piecewise polynomials in the approximation spaces.

Proof: Lemma 5.9 proves (5.60). We have proved

$$\sum_{j \in \mathbb{Z}_{w(i)}} |\langle \ell_{i'j'}, w_{ij} \rangle| \le \frac{81}{28}, \quad (i,j), (i',j') \in \mathbb{U}, i > i'$$

in Lemma 3.2. In view of Lemma 5.1 and Lemma 5.2, we have

$$\langle \ell_{i'j'}^{(k)}, w_{ij}^{(k)} \rangle = \langle \ell_{i'j'}, w_{ij} \rangle$$

Thus

$$\sum_{j \in \mathbb{Z}_{w(i)}} |\langle \ell_{i'j'}^{(k)}, w_{ij}^{(k)} \rangle| \le \frac{81}{28}, \quad (i,j), (i',j') \in \mathbb{U}, i > i'.$$
(5.63)

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Direct computation shows (5.62).

Lemma 5.12 The constructed wavelets and collocation functionals have vanishing moment.

Proof: From Lemma 5.7 and Lemma 5.8, we obtain

$$\langle \ell_{i'j'}^{(k)}, w_{0j}^{(k)} \rangle = \langle \ell_{1l'}^{(k)}, w_{0j}^{(k)} \circ \phi_{k,\mathbf{e}'} \rangle = 0, \quad j \in \mathbb{Z}_3,$$
(5.64)

where, $i' \geq 2$, $j' \in \mathbb{Z}_{w(i')}$ and $j' = 9\mu(\mathbf{e}') + l'$, $l' \in \mathbb{Z}_9$ with $\mathbf{e}' \in \mathbb{Z}_4^{i'-1}$. We now consider $\mathcal{T}_{k,\mathbf{e}}$ as an operator from $L^2(\triangle_k)$ to $L^2(\triangle_k)$ and Let $\mathcal{T}_{k,\mathbf{e}}^*$ be the adjoint of $\mathcal{T}_{k,\mathbf{e}}$. It is easy to see that for $y \in L^2(\triangle_k)$

$$\mathcal{T}_{k,\mathbf{e}}^* y = J_{\phi_{k,\mathbf{e}}} y \circ \phi_{k,\mathbf{e}},\tag{5.65}$$

where $J_{\phi_{k,\mathbf{e}}}$ is the Jacobian of mapping $\phi_{k,\mathbf{e}}$. Therefore, we have that

$$(w_{ij}^{(k)}, w_{0j'}^{(k)}) = (\mathcal{T}_{k,\mathbf{e}} w_{1l}^{(k)}, w_{0j'}^{(k)}) = (w_{1l}^{(k)}, \mathcal{T}_{k,\mathbf{e}}^* w_{0j'}^{(k)}) = 0,$$

where, $i \geq 2$, $j \in \mathbb{Z}_{w(i)}$ and $j = 9\mu(\mathbf{e}) + l$, $l \in \mathbb{Z}_9$ with $\mathbf{e} \in \mathbb{Z}_4^{i-1}$. The last equality holds because $\mathcal{T}_{k,\mathbf{e}}^* w_{0j'}^{(k)}$ is a polynomial of total degree $\leq s - 1 = 1$, with s = 2 and $w_{1l}^{(k)}$ satisfies Lemma 5.5 and Lemma 5.6. Thus for any polynomials $p \in \pi_2$,

$$\langle \ell_{ij}^{(k)}, p \rangle = 0, \quad (w_{ij}^k, p) = 0, \quad (i, j) \in \mathbb{U}, \ i \ge 1.$$
 (5.66)

Lemma 5.13 Both basis functions and collocation functionals are uniformly bounded.

Proof: For $(i, j) \in \mathbb{U}$, $i \ge 2$, $j = \mu(\mathbf{e})r + l$, $l \in \mathbb{Z}_9$, Lemma 5.1 and Lemma 3.4 yields

$$|\langle \ell_{ij}^{(k)}, v \rangle| = |\langle \ell_{ij}, v \circ T_k \rangle| = |\langle \ell_{1l}, v \circ T_k \circ \phi_{\mathbf{e}} \rangle| \le \|\mathbf{C}'\|_{\infty} \|v\|_{\infty}$$

and Lemma 5.2 yields

$$\|w_{ij}^{(k)}\|_{\infty} = \|w_{ij}T_k^{-1}\|_{\infty} \le \|\mathbf{C}\|_{\infty} \max_{j \in \mathbb{Z}_{12}} \|\psi_j\|_{\infty},$$

The last inequality holds because of Lemma 3.4.

Lemma 5.14 The dimension of spaces \mathbb{F}_n and \mathbb{W}_n grows exponentially in n and the diameters d_n decay exponentially in n too, where

$$d_n := \max\{ d\left(S_{nm}^{(k)}\right), \ (n,m) \in \mathbb{U}, k \in \mathbb{Z}_p \}.$$

Proof: dim $\mathbb{W}_0 = 3p$, dim $\mathbb{F}_n = 3p.4^n$, $n \in \mathbb{Z}_0$ and dim $\mathbb{W}_n = 9p.4^{n-1}$, $n \ge 1$, $n \in \mathbb{Z}$. Let $\hat{d} = \max\{d^{(k)}, k \in \mathbb{Z}_p\}$, where $d^{(k)}$ is the diameter of each triangle Δ_k , $k \in \mathbb{Z}_p$ in the triangulation $\Delta_{\mathbf{p}}$. For the diameter we have $d_0 = \hat{d}$ and $d_n \le c \cdot \hat{d}_{\frac{1}{2^{n-1}}} = c2^{-n}$, where c > 0 is some constant.

Lemma 5.15 There exists constant θ_2 and θ_3 such that for all $n \in \mathbb{N}_0$ and v having form

$$v := \sum_{(i,j)\in\mathbb{U}_n} v_{k,ij} w_{ij}^{(k)}, \quad k \in \mathbb{Z}_p,$$
(5.67)

$$\theta_2 \|\mathbf{v}\|_{\infty} \le \|v\|_{\infty} \le \theta_3(n+1) \|\mathbf{E}_n \mathbf{v}\|_{\infty},\tag{5.68}$$

where $\mathbf{v} := [v_{k,ij} : (i,j) \in \mathbb{U}_n, \ k \in \mathbb{Z}_p]^T$.

Proof: Note the constructed wavelets $w_{ij}^{(k)}$ and collocation functionals $\ell_{i'j'}^{(k)}$, with $(i, j), (i', j') \in \mathbb{U}, k \in \mathbb{Z}_p$ have properties stated in Lemma 5.10 - Lemma 5.13, the proof is then similar to the proof of Lemma 3.6.

Lemma 5.16 Let \mathcal{P}_n be the projection operator from \mathbb{X} onto \mathbb{F}_n defined by

$$\langle \ell_{ij}^{(k)}, \mathcal{P}_n x \rangle = \langle \ell_{ij}^{(k)}, x \rangle, (i, j) \in \mathbb{U}_n.$$
 (5.69)

For $v \in L^{\infty}(E)$ we set

$$\mathcal{P}_n v = \sum_{(i,j) \in \mathbb{U}_n} v_{k,ij} w_{ij}^{(k)}, \quad k \in \mathbb{Z}_p.$$

Then, for any $x \in L^{\infty}(E)$, there holds

$$\lim_{n \to \infty} \|\mathcal{P}_n x - x\|_{\infty} = 0.$$

Proof: The proof is similar to that of Lemma 3.7.

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Lemma 5.17 There exists a positive constant c such that for $u \in W^{k,\infty}(E)$

$$\operatorname{dist}(u, \mathbb{F}_n) \le c\mu^{-kn/2} \|u\|_{k,\infty}.$$
(5.70)

Proof: The proof is similar to that of Lemma 3.8.

Theorem 5.18 If we use the constructed wavelet basis $w_{ij}^{(k)}$ and collocation functionals $\ell_{ij}^{(k)}$ on a polygon solve (2.2) with the spacial domain being a polygon, then Theorem 4.13, Theorem 4.14, Theorem 4.15, and Theorem 4.16 continue to hold.

Proof: The theorem follows from the properties Lemma 5.10 through 5.17 of the constructed wavelet basis functions and collocation functionals over polygons. \Box

Chapter 6

Numerical Experiments for Solving Integral Equations

In this chapter, we use methods proposed in this thesis to solve several typical examples. As far as compression strategies are concerned, we use both distance computing strategy 2 and block truncation strategy. The block truncation strategy manifests good truncation effect in terms of presenting satisfactory results. The resulting linear systems are solved by Gauss-Seidel iterative scheme developed in Section 4.8. When level n is relatively large (say $n \ge 6$), the iterative scheme dramatically reduce the computing time by comparing with Gaussian elimination method. When the matrix entries are computed numerically, the quadrature rule presented in Section 4.3 is adopted. All the computation are done using parallel computing algorithm, which is discussed in Section 6.4.

6.1 On the Unit Triangle

Consider following integral equation of the second kind

$$u(x) - \int_{\Omega} \frac{1}{|x-y|} u(y) \mathrm{d}y = f(x), \tag{6.1}$$

_	-		-					
		without quadrature		with quadrature				
	n	L^{∞} -error	order	(m_1,m_2)	L^{∞} -error	order	compression rate	
	1	1.513187e-1		(40, 40)	1.513184e-1		1.000000	
	2	3.599605e-2	2.0717	(40, 40)	3.599379e-2	2.0718	1.000000	
	(3,0)	7.499082e-3	2.2631	(80, 40)	7.498160e-3	2.2631	1.000000	
	(3,1)	1.849142e-3	2.0199	(80, 40)	1.849507e-3	2.0194	0.526763	
	(3,2)	4.644798e-4	1.9932	(80, 40)	4.646119e-4	1.9930	0.228013	
	(3,3)	1.168886e-4	1.9905	(80, 40)	1.168323e-4	1.9916	0.085760	
	(3,4)	2.955001e-5	1.9839	(80, 40)	2.982907e-5	1.9696	0.030163	
	(3,5)	1.081540e-5	1.4501	(80, 40)	1.082801e-5	1.4620	0.010460	

Compression strategy: strategy 2 with $\nu = 1$

Table 6.1: Comparison of compressed solutions using multilevel iteration method, with and without quadrature rule, number of iteration=15.

where $x, y \in \mathbb{R}^2$, $x = (x_1, x_2)$, $y = (y_1, y_2)$, and Ω is the unit triangle.

We choose

$$f(x) = x_1^2 + x_2^2 - \int_{\Omega} \frac{y_1^2 + y_2^2}{|x - y|} dy$$
(6.2)

so that the exact solution to (6.1) is

$$u(x) = x_1^2 + x_2^2. ag{6.3}$$

Example 6.1 In this example, we apply compression strategy 2 and use Gauss-Seidel type multilevel iteration method to solve the resulting linear system. The initial level k used for the multilevel iteration is chosen as k = 3. Note when the level $n = k + \ell = 3 + 5 = 8$, the matrix size is 196608 × 196608, and the compression rate is about 1%. It's surprising that even the matrix entries are computed without quadrature (computed by analytical formula. Only when the domain of integration is unit triangle, can we find the explicit formula by Mathematica. When the domain of integration is a general triangle, no explicit formula is available), for n = 8, the result is far from satisfactory one. This shows the limitation of strategy 2.

	without qua	drature	with quadrature							
n	L^{∞} -error	order	(m_1,m_2)	L^{∞} -error	order	compression rate				
1	1.513187e-1		(40, 40)	1.513184e-1		1.000000				
2	3.599605e-2	2.0717	(40, 40)	3.599379e-2	2.0718	1.000000				
(3,0)	7.499082e-3	2.2631	(80, 40)	7.499123e-3	2.2630	1.000000				
(3,1)	1.850492e-3	2.0188	(80, 40)	1.850509e-3	2.0188	0.431732				
(3,2)	4.648565e-4	1.9931	(80, 40)	4.648294e-4	1.9931	0.181965				
(3,3)	1.171548e-4	1.9884	(80, 40)	1.174169e-4	1.9851	0.067760				
(3,4)	2.999366e-5	1.9657	(80, 40)	2.980252e-5	1.9781	0.023004				
(3,5)	8.010510e-6	1.9047	(80, 40)	7.920534e-6	1.9118	0.007332				

Compression strategy: block truncation with $p_{i'i} = \sqrt{3}$

Table 6.2: Comparison of compressed solutions using multilevel iteration method, with and without quadrature rule, number of iteration=15.

Example 6.2 In this example, we apply block truncation strategy and use Gauss-Seidel type multilevel iteration method to solve the resulting linear system. The initial level k used for the multilevel iteration is chosen as k = 3. The block truncation parameter $p_{i'i} = \sqrt{3}$, $i', i \in \mathbb{Z}_{n+1}$. Note compression rates in this case are even smaller than their counterparts using strategy 2. What's most amazing is that when n = 8, the block truncation scheme continues to give satisfied solution. This suggests block truncation scheme is superior to strategy 2 in order to catch critical elements.

6.2 On General Triangles

We consider the triangle \triangle with three vertices (0,0), (1,0) and (0.5,1). As before, we take the kernel

$$K(x, y, \xi, \eta) = \frac{1}{\sqrt{(x-\xi)^2 + (y-\eta)^2}}, \quad (x, y), (\xi, \eta) \in \Delta,$$
(6.4)

and

$$f(x,y) = x^{2} + y^{2} - \int_{\Delta} \frac{\xi^{2} + \eta^{2}}{\sqrt{(x-\xi)^{2} + (y-\eta)^{2}}} \mathrm{d}\xi \mathrm{d}\eta, \quad (x,y) \in \Delta$$
(6.5)

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		8	<u> </u>	00				
	full collocation	on solution u_n	compressed solution \tilde{u}_n					
n	L^{∞} -error	order	(m_1,m_2)	L^{∞} -error	order	compression rate		
1	1.100676e-1		(40, 40)	1.100676e-1		1.000000		
(2,0)	2.081240e-2	2.4029	(80, 40)	2.081280e-2	2.4028	1.000000		
(2,1)	4.972866e-3	2.0653	(80, 40)	4.973775e-3	2.0651	0.894531		
(2,2)	1.285098e-3	1.9522	(80, 40)	1.286323e-3	1.9511	0.520172		
(2,3)	3.230465e-4	1.9921	(80, 40)	3.232655e-4	1.9925	0.227601		
(2,4)			(80, 40)	7.800080e-5	2.0512	0.085734		
(2,5)			(80, 40)	2.100672e-5	1.8926	0.030162		
(2,6)			(80, 40)	5.914368e-6	1.8286	0.010460		

Results for the general triangle \triangle : strategy 2 with $\nu = 1$

Table 6.3: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 2, on a general triangle).

so that $u(x,y) = x^2 + y^2$, $(x,y) \in \Delta$ is the exact solution to equation (5.9). In this case, no analytical formula is available for computing the matrix entry. All entries are computed by quadrature method proposed in Section 4.3. For the uncompressed solution, the linear system is solved by stand Gaussian elimination method, and the number of subdivision mused in the quadrature is taken as m = 40. While, for the compressed solution, the linear system is solved by Gauss-Seidel type iterative method presented in Section 4.8 (number of iteration=20), and the number of subdivision used in quadrature is as shown in Table 6.3 and Table 6.4.

Example 6.3 In this example, we use the truncation strategy 2 with $\nu = 1$. The computational results are listed in Table 6.3.

Example 6.4 In this example, we use the block truncation strategy with parameters $p_{i'i} = \sqrt{3}$. The results are listed in Table 6.4.

	full collocation	on solution u_n	compressed solution \tilde{u}_n								
n	L^{∞} -error	order	(m_1,m_2)	L^{∞} -error	order	compression rate					
1	1.100676e-1		(40, 40)	1.100676e-1		1.000000					
(2,0)	2.081240e-2	2.4029	(80, 40)	2.081280e-2	2.4028	1.000000					
(2,1)	4.972866e-3	2.0653	(80, 40)	4.974875e-3	2.0647	0.758301					
(2,2)	1.285098e-3	1.9522	(80, 40)	1.285873e-3	1.9519	0.416626					
(2,3)	3.230465e-4	1.9921	(80, 40)	3.257940e-4	1.9807	0.181021					
(2,4)			(80, 40)	7.869679e-5	2.0496	0.067701					
(2,5)			(80, 40)	2.195469e-5	1.8418	0.023000					
(2,6)			(80, 40)	7.829137e-6	1.4876	0.007331					

Results for the general triangle \triangle : block truncation with $p_{i'i} = \sqrt{3}$

Table 6.4: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (block truncation, on a general triangle).

6.3 On Polygons

In the following examples, as before, we take kernel as $K(x, y, \xi, \eta) = \frac{1}{\sqrt{(x-\xi)^2+(y-\eta)^2}}$, and $f(x,y) = x^2 + y^2 - \int_E \frac{\xi^2 + \eta^2}{\sqrt{(x-\xi)^2+(y-\eta)^2}} d\xi d\eta$ so that $u(x) = x^2 + y^2$, where $(x,y) \in E$, is the exact solution to (5.44). For the uncompressed solution, the linear system is solved by stand Gaussian elimination method, and the number of subdivision m_n used in the quadrature is taken as $m_n = 40$. While, for the compressed solution, the linear system is solved by Gauss-Seidel type iterative method presented in Section 4.8 (number of iteration=20), and the number of subdivision used in quadrature is $m_n = 80$. In Example 6.5 and Example 6.6, we consider the quadrangle E with four vertices (0,0), (1,0), (2,1.2), (0.5,1) (see Fig. 6.1).

Example 6.5 In this example, for the compressed solution, the truncation strategy 2 with $\nu = 1$ is applied. The computational results are listed in Table 6.5.

Example 6.6 In this example, for the compressed solution, the block truncation strategy with parameters $p_{i'i} = \sqrt{3}$ is applied. The computational results are listed in Table 6.6.

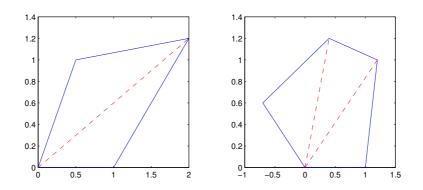


Figure 6.1: Computational domains of Example 6.5 and Example 6.7

Results for the integral equation ove	r a quadrangle: strategy 2 with $\nu = 1$
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	full collocation solution u_n			cor	npressed solut		
n	m_n	L^{∞} -error	order	m_n	L^{∞} -error	order	compression rate
1	40	3.607234e-1		40	3.607234e-1		1.000000
(2,0)	40	9.443242e-2	1.9335	80	9.443194e-2	1.9335	1.000000
(3,0)	40	2.282575e-2	2.0486	80	2.282116e-2	2.0489	0.894531
(4,0)	40	5.735749e-3	1.9926	80	5.753102e-3	1.9880	0.520172
(5,0)	40	1.456519e-3	1.9775	80	1.449515e-3	1.9888	0.227601
(6,0)				80	4.190603e-4	1.7903	0.085734

Table 6.5: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 2, on a quadrangle).

	full collocation solution u_n			cor	npressed solut		
n	m_n	L^{∞} -error	order	m_n	L^{∞} -error	order	compression rate
1	40	3.607234e-1		40	3.607234e-1		1.000000
(2,0)	40	9.443242e-2	1.9335	80	9.443194e-2	1.9335	1.000000
(3,0)	40	2.282575e-2	2.0486	80	2.272426e-2	2.0550	0.758301
(4,0)	40	5.735749e-3	1.9926	80	5.891051e-3	1.9476	0.416626
(5,0)	40	1.456519e-3	1.9775	80	1.492354e-3	1.9809	0.181021
(6,0)				80	9.469365e-4	0.6563	0.067701

Results for the integral equation over a quadrangle: block truncation with $p_{i'i} = \sqrt{3}$

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Table 6.6: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (block truncation, on a quadrangle).

In Example 6.7 and Example 6.8, we consider the pentagon E with five vertices (0,0), (1,0), (1.2,1), (0.4,1.2), (-0.7,0.6) (see Fig. 6.1).

Example 6.7 In this example, for the compressed solution, the truncation strategy 2 with $\nu = 1$ is applied. The computational results are listed in Table 6.7.

Example 6.8 In this example, for the compressed solution, the block truncation strategy with parameters $p_{i'i} = \sqrt{3}$ is applied. The computational results are listed in Table 6.8.

6.4 Parallel Computing

As we see, when the level of the resolution is higher (for our case, when n > 7), even a sparse storage scheme is adopted, due to the fast increasing of nonzero elements, the computation can not be fulfilled on one stand alone computer. The problem encountered is the limit of the physical memory of the available machine. The other problem is the painful lengthy computing time when the level is getting higher. Therefore if we want to compute higher level's solution, one definitely good answer is to appeal to parallel computing.

To achieve parallel computing using methods proposed in previous chapters, one must con-

	The state of the model at equation over a period gon. Strategy 2 with $\nu = 1$									
		full collocation solution u_n			compressed solution \tilde{u}_n					
n	m_n	L^{∞} -error	order	m_n	L^{∞} -error	order	compression rate			
1	40	2.015684e-1			2.0157e-1		1.000000			
(2,0)	40	4.076680e-2	2.3058	80	4.076633e-2	2.3058	1.000000			
(2,1)	40	1.017880e-2	2.0018	80	1.017623e-2	2.0022	0.894531			
(2,2)	40	2.593854e-3	1.9724	80	2.596571e-3	1.9705	0.520172			
(2,3)	40	6.440248e-4	2.0099	80	6.434862e-4	2.0126	0.227601			
(2,4)				80	1.587823e-4	2.0189	0.085734			
(2,5)				80	5.511457e-5	1.5265	0.030162			

Results for the integral equation over a pentagon: strategy 2 with $\nu = 1$

Table 6.7: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (strategy 2, on a pentagon).

	Testitis for the model of equation over a periodon. Since transation with $p_{i'i} = \sqrt{5}$									
		full collocation solution u_n			compressed solution \tilde{u}_n					
n	m_n	L^{∞} -error	order	m_n	L^{∞} -error	order	compression rate			
1	40	2.015684e-1			2.0157e-1	40	1.000000			
(2,0)	40	4.076680e-2	2.3058	80	4.076633e-2	2.3058	1.000000			
(2,1)	40	1.017880e-2	2.0018	80	1.019395e-2	1.9997	0.758301			
(2,2)	40	2.593854e-3	1.9724	80	2.624460e-3	1.9576	0.416626			
(2,3)	40	6.440248e-4	2.0099	80	6.675855e-4	1.9750	0.181021			
(2,4)				80	1.939918e-4	1.7830	0.067701			
(2,5)				80	9.054316e-5	1.0993	0.023000			

Results for the integral equation over a pentagon: block truncation with $p_{i'i}=\sqrt{3}$

Table 6.8: Comparison of solutions obtained by the full collocation method and by our compressed collocation method (block truncation, on a pentagon). sider two factors:

- to be able to simultaneously generate the coefficient matrix of the discrete linear system, henceforth the coefficient matrix needs to be partitioned into pieces distributed on different machines
- to be able to simultaneously solve the resulting linear system using the distributed matrix on different machines

It turns out that we can easily design our algorithm, taking advantage of both sparse storage and parallel computing. The main idea is that we will partition the coefficient matrix along row direction. Thus on different machines, disjoint partitions of the matrix are simultaneously computed. To solve the linear system, we adopt the Gauss-Seidel type multilevel iteration methods proposed in Section 4.8. The algorithm is as follows:

A Parallel computing algorithm:

1. Allocate the computing loads roughly evenly to different computers. Let $np \ge 2$ be the number of processes. Compute the quota of rows

$$Q = \left\lfloor \frac{f(n)}{np} \right\rfloor \tag{6.6}$$

where, as before f(n) is the dimension of the approximation space \mathbb{F}_n , i.e., the order of the matrix $\mathbf{A}_n = \mathbf{K}_n - \mathbf{E}_n$. Remember we are trying to generate and solve the linear system (2.18) or (4.102). Recall we denote elements of \mathbf{K}_n by $K_{i'j',ij}$, elements of \mathbf{E}_n by $E_{i'j',ij}$, and elements of \mathbf{f}_n by $f_{i'j'}$, where $(i', j'), (i, j) \in \mathbb{U}_n$. Let r_i be the number of rows to be allocated for process $i = 0, 1, \dots, np - 1$, then

$$r_i = Q$$
, for $i \in \mathbb{Z}_{np-1}$ and $r_{np-1} = f(n) - \sum_{i=0}^{np-2} r_i$; (6.7)

2. Simultaneously compute nonzero coefficient elements $K_{i'j',ij}$ and $E_{i'j',ij}$ (by applying matrix compression) and the right handed vector \mathbf{f}_n on each distributed machines. No communication between the computing nodes is needed at this step.

As we know, any element $K_{i'j',ij}$ and $E_{i'j',ij}$ are completely determined by its index

(i'j', ij). Denote the algorithm to compute $K_{i'j',ij}$ by $\mathbf{Kelt}(i', j', i, j)$, the algorithm to compute $E_{i'j',ij}$ by $\mathbf{Eelt}(i', j', i, j)$, and the algorithm to compute $f_{i'j'}$ by $\mathbf{felt}(i', j')$. Note differences between *local* indices of matrix elements on each machine and the global indices of the elements in the original matrix. All algorithms $\mathbf{Kelt}(i', j', i, j)$, $\mathbf{Eelt}(i', j', i, j)$ and $\mathbf{felt}(i', j')$ assumes the global indices.

- 3. Employ the Gauss-Seidel multilevel iteration method (see Section 4.8) to solve the linear system simultaneously on different machines. Communications between nodes are assumed.
 - (a) Obtain the initial solution \mathbf{x}_0 on process 0. Then Broadcast \mathbf{x}_0 to all other processes $1, 2, \dots, np-1$.
 - (b) On each nodes, simultaneously do matrix-vector multiplication to obtain a locally updated partial solution $\mathbf{x}_{\nu}^{(i)}$, where, *i* represents the index of process, $i \in \mathbb{Z}_{np}$, and ν stands for the number of iteration, and $\nu > 1$. Let $\mathbf{A}^{(i)} = [A_{jk}^{(i)} : j \in \mathbb{Z}_{r_i}, k \in \mathbb{Z}_{f(n)}, i \in \mathbb{Z}_{np}]$ be the partial matrix computed on process *i*, and $\mathbf{x}_{\nu}^{(i)} = [x_{\nu}^{(i)}(j) : j \in \mathbb{Z}_{f(n)}, i \in \mathbb{Z}_{np}]$ be the local solutions on process *i*, and $\mathbf{f}^{(i)} = [f_j^{(i)} : j \in \mathbb{Z}_{r_i}]$. Then for each $\nu > 1$,

$$x_{\nu}^{(i)}(j) = f_j^{(i)} + \sum_{k=0}^{f(n)-1} A_{jk}^{(i)} x_{\nu-1}(k), \quad j \in \mathbb{Z}_{r_i},$$
(6.8)

where $i \in \mathbb{Z}_{np}$, and $\mathbf{x}_{\nu-1} = [x_{\nu-1}(k) : k \in \mathbb{Z}_{f(n)}]$ is the previous computed global solution.

- (c) Assemble the updated solution \mathbf{x}_{ν} on node 0 from the local partial segments $\mathbf{x}_{\nu}^{(i)}, i \in \mathbb{Z}_{np}$ and compute the current approximation error.
- (d) Broadcast the newest solution \mathbf{x}_{ν} to all other nodes in the current communicator.
- (e) Do step 3b-step 3d again until satisfied solution is achieved.

All the examples using above parallel algorithm are computed on energy.cluster.wvu.edu. The cluster has one frontend node and 27 computing nodes. Each node has 1.2 Ghz Pentium, Dual processors, 1 Gbyte of memory shared by the two processors and 18 Gbyte of disk and has a 100 Mb ethernet interconnect. **Example 6.9** To demonstrate the ideas proposed in the above algorithm, we solve the integral equation (6.1) again using parallel algorithm. For the example of n = 7, compression is assumed. While for the example of n = 5, no compression is assumed. The linear system is solved by Gauss-Seidel type multilevel iteration method. All the entries are computed by analytical formula. We tabulate in Table 6.9 the computing time for different number of processes np. t_m is the average time consumed by each process to generate the sparse coefficient matrix. t_s is the average time spent by each process to solving the linear system by Gauss-Seidel type multilevel iteration method. When n = 7, the computing requires the same number of computing nodes as number of processes np. While when n = 5, we require all the two processors on each node.

The results are also plotted in Figure 6.2 and Figure 6.3. From both figures, we clearly see, as the number of processes increase, the average time t_m needed for each process to generate the partial matrix is decreasing. As described in the parallel computing algorithm, no communication between nodes is assumed for this purpose. While, the average time needed for each process to solve the linear system by iteration method (matrix-vector multiplication) is more or less the same as when there is only one process (the serial computing). This is as expected. The reason is as follows:

The matrix-vector multiplication time needed in each process only differs by a very insignificant factor when np increases. While more and more communications among processes are caused due to more processes are introduced into the whole communicator. The increased communication time more or less offset the reduced time for matrix-vector multiplication. Therefore, almost equal time for solving the system observed when different number of processes are required. The fluctuation of each curve is apparently due to the dynamical load of the cluster. Since a number of other jobs are also simultaneously running in the cluster. I imagine that if only one job is running in the cluster before it's done, then the fluctuation won't be observed.

	level $n =$	= 7	level <i>i</i>	n = 5
$^{\mathrm{np}}$	$t_m(sec)$	$t_s(sec)$	$t_m(\text{sec})$	$t_s(sec)$
1	14023	1349	917	103
2	5101	1345	687	140
3	2934	1317	406	134
4	2258	2611	229	89
5	2335	2122	256	131
6	1282	1289	204	136
7	1069	1283	205	172
8	891	1281	143	85
9	793	1352	148	85
10	713	1275	120	127
11	648	1274	121	127
12	594	1272	102	127
13	548	1271	92	84
14	509	1270	84	84
15	539	1902	85	84
16	471	1268	71	84
17	443	1266	73	87
18	418	1267	59	167
19	440	1894	62	84
20	415	1892	50	84
21	367	1264		
22	375	1890		
23	361	1891		
24	321	1263		

Table 6.9: Comparison of times consumed when number of processes varies.

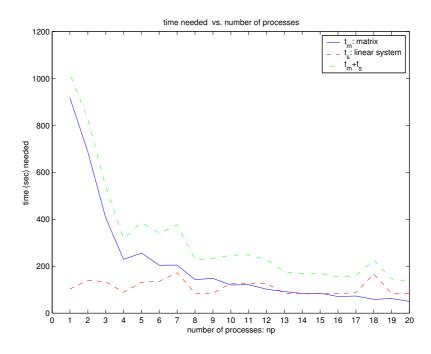


Figure 6.2: Time needed vs. number of processes: level n = 5

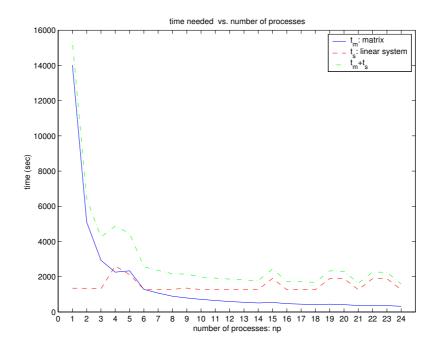


Figure 6.3: Time needed vs. number of processes: level n = 7

Bibliography

- [Al] B.K.ALPERT, A class of bases in L² for the sparse representation of integral operators, SIAM J. Math. Anal., 24 (1993), pp. 246-262
- [An] P.M. ANSELONE, Collectively Compact Operator Approximation Theorey and Applications to Integral Equations, Prentice-Hall, Englewood Cliffs, NJ, 1971
- [At1] K. E. ATKINSON, A survey of boundary integral equation methods for the numerical solution of Laplace's equation in three dimensions, in Numerical Solution of Integral Equations, M. Golberg, ed., Plenum Press, NY 1990, pp. 1-34
- [At2] K. E. ATKINSON, The Numerical Solution of Integral Equations of Second Kind, Cambridge University Press, Cambridge, 1997
- [AC1] K.E. ATKINSON AND D. CHIEN, Piecewise polynomial collocation for boundary integral equations, SIAM J. Sci. Comput., 16 (1994), pp. 651-681
- [AC2] K.E. ATKINSON AND D. CHIEN, A fast matrix-vector multiplication method for solving the radiosity equation, Adv. Comput. Math., 12 (2000), pp. 151-174
- [ACr] K.E. ATKINSON AND G. CHANDLER, The collocation method for solving the radiosity equation for unoccluded surfaces, J. Integral Equations Appl., 10 (1998), pp. 253-290
- [AGS] K. E. ATKINSON, I. GRAHAM AND I. SLOAN, Piecewise continuous collocation for integral equations, SIAM J. Numer. Anal., 20 (1983), 172-186

- [BCR] G. BEYLKIN, R. COIFMAN AND V. ROKHLIN, Fast wavelet transforms and numerical algorithms I, Comm. Pure Appl. Math., 44 (1991), pp. 141-183
- [CMX1] Z. CHEN, C. A. MICCHELLI AND Y. XU, The Petrov-Galerkin methods for second kind integral equations II: Multiwavelet scheme, Adv. Comput. Math., 7 (1997), pp. 199-233
- [CMX2] Z. CHEN, C. A. MICCHELLI AND Y. XU, A construction of interpolating wavelets on invariant sets, Math. Comp., 68 (1999), pp. 1569-1587
- [CMX3] Z. CHEN, C. A. MICCHELLI AND Y. XU, Fast Collocation Methods for Second Kind Integral Equations, SIAM J. NUMER. ANAL., Vol. 40, No. 1, pp. 344-375
- [CMX4] Z. CHEN, C. A. MICCHELLI AND Y. XU, A multilevel Method for Solving Operator Equations, J. Math. Anal. and Appl. 262, 688-699 (2001)
- [CMX5] Z. CHEN, C. A. MICCHELLI AND Y. XU, Discrete wavelet Petrov-Galerkin methods, Adv. Comput. Math., 16 (2002), pp. 1-28
- [CMX6] Z. CHEN, C. A. MICCHELLI AND Y. XU, A multilevel Method for Solving Operator Equations, J. Math. Anal. and Appl. 262, 688-699 (2001)
- [CWX1] Z. CHEN, B. WU, AND Y. XU, Multilevel augmentation methods for solving operator equations, submitted
- [CWX2] Z. CHEN, B. WU AND Y. XU, An Implementation of Fast Collocation Methods for Integral Equations, preprint
- [CX] Z. CHEN AND Y. XU, The Petrov-Galerkin and iterated Petrov-Galerkin methods, Adv. Comput. Math., 16 (2002), pp. 1-28
- [DKPS] W. DAHMEN, B. KLEEMANN, S. PRÖSSDORF AND R. SCHNEIDER, A multiscale method for the double layer potential equation on a polyhedron, in Advances in Computational Mathematics: New Delhi, India, eds. H. P. Dikshit and C. A. Micchelli, World Scientific, Singapore, 1994.

- [DPS1] W. DAHMEM, S. PROESSDORF, AND R. SCHINEIDER, Wavelet approximation methods for pseudodifferential equations II: matrix compression and fast solutions, Adv. Comput. Math., 1 (1993), pp. 259-335
- [DPS2] W. DAHMEN, S. PRÖSSDORF AND R. SCHNEIDER, Wavelet approximation methods for psuedodifferential equations II: Matrix compression and fast solutions, Adv. Comput. Math., 1 (1993), 259-335.
- [FMX] W. FANG, F. MA AND Y. XU, Multilevel Iteration Methods for solving Integral Equations of the Second Kind, J. Integral Equations Appl., (to appear)
- [FWX] W. FANG, Y. WANG AND Y. XU, An Implementation of Fast Wavelet Galerkin Methods for Integral Equations of the Second Kind, J. Scientific Computing (to appear)
- [H] J.E. HUTCHINSON, Fractals and self similarity, India Univ. Math. J. 30 (1981), 713-747.
- [K] R. KRESS, *Linear Integral Equations*, 2nd ed., Springer-Verlag, New York, 1999
- [KX] H. KANEKO AND Y. XU, Gauss-Type Quadratures for Weakly Singular Integrals and Their Application to Fredholm Integral Equations of the Second Kind, Math. Comp. 62 (1994) 739-753
- [MSX1] C. A. MICCHELLI, T. SAUER AND Y. XU, A construction of refinable sets for interpolating wavelets, Result. Math. 34 (1998) 359-372
- [MSX2] C. A. MICCHELLI, T. SAUER AND Y. XU, Subdivision schemes for iterated function systems,, Proc. Amer. Math. Socl, 129 (2001), pp. 1861-1872
- [MX1] C. A. MICCHELLI, AND Y. XU, Using the matrix refinement equation for the construction of wavelets on invariant sets, Appl, Comput. Harmon. Anal., 1 (1994), pp. 391-401

- [MX2] C. A. MICCHELLI, AND Y. XU, Reconstruction and decomposition algorithms for biorthogonal multiwavelets, Multidimen. Systems Signal Process., 8 (1997), pp. 31-69
- [MXZ] C. A. MICCHELLI, Y. XU AND Y. ZHAO, Wavelet Garlerkin Methods for Second-Kind Integral Equations, J. Comp. and Appl. Math. 86 (1997) 251-270
- [PS] T. VON PETERSDORFF AND C. SCHWAB, Wavelet approximation of first kind integral equations in a polygon, Numer. Math., 74 (1996), pp. 479-516
- [PSS] T. VON PETERSDORFF, C. SCHWAB, AND R. SCHNEIDER, Multiwavelets for second-kind integral equations, SIAM J. Numer. Anal., 34 (1997), pp. 2212-2227
- [R] A. RATHSFELD, A wavelet algorithm for the solution of a singular integral equaiton over a smooth two-dimensional manifold, J. Integral Equaitons Appl., 10 (1998), pp. 445-501
- [Ri] J. R. RICE, On the degree of convergence of nonlinear spline approximaton, Approx. with Emphasis on Spline Functions (I.J. Schoenberg, ed.), Academic Press, New York, 1969, pp. 349-365
- [XZ] Y. XU AND Y. ZHAO, An Extrapolation Method for a Class of Boundary Integral Equations, Math. Compu. 65 (1996), 587-610