A Collocation Method for
Fredholm Integral Equations of the Second Kind

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Abstract. An interpolation scheme based on piecewise cubic polynomials with Gaussian points as interpolation points is analyzed and applied to the solution of Fredholm equations of the second kind.

Introduction. We consider an interpolation scheme based on piecewise polynomials with continuous first derivatives and the Gaussian points as interpolation points. A collocation method based on this interpolation scheme is applied to one- and two-dimensional Fredholm equations of the second kind.

This scheme has been applied to a collocation method by de Boor and Swartz [4] and Houstis [9] for the numerical solution of ordinary differential equations. Also, Douglas and Dupont [6], [7], [8] and Houstis [10], [11] have studied a collocation method for partial differential equations based on the above scheme. A survey of numerical methods for the solution of Fredholm integral equations of the second kind is given by Atkinson [1].

In Part I we present the formulation and error analysis of the interpolation scheme. In Part II we apply this scheme to the solution of Fredholm integral equations of the second kind and give an experimental comparison with Nyström’s method.

I. Piecewise Cubic Hermite Interpolation at the Gaussian Points

1. One-Dimensional Interpolation Scheme. Let \( \Delta = (x_j)_{j=1}^{N+1} \) be a partition of \( I = [a, b] \), \( h_i \equiv x_{i+1} - x_i \), \( I_i = [x_{i+1}, x_{i+2}] \) and \( h = \max h_i \). Throughout this report we denote by \( P_3 \) the set of polynomials of degree less than 4, and \( P_{3, \Delta} \) the set of functions that reduce to polynomials of degree less than 4 in each subinterval \( [x_i, x_{i+1}] \). Also, we denote by \( H_\Delta \) the \((2N + 2)\)-dimensional vector space of all continuously differentiable piecewise cubic polynomials with respect to \( \Delta \).

The Gaussian points in the subinterval \( [x_i, x_{i+1}] \) are

\[
\xi_{2j+i} = \frac{x_j + x_{j+1}}{2} + \frac{(-1)^i h_j}{\sqrt{3}}, \quad i = 1, 2.
\]

Let \( E(I) \) be the space of real-valued functions defined on \( I \). We introduce an interpolation operator \( Q_N : E(I) \rightarrow H_\Delta \) such that
(1.2) \((Q_N f)(\sigma_j) = f(\sigma_j), \quad l = 1, \ldots, 2N + 2,\)

where \(\sigma_1 = a, \sigma_l = \xi_{2l+l}, j = 1, \ldots, N, i = 1, 2, \sigma_{2N+2} = b.\)

This interpolation scheme is well defined. In fact, if \(h(x) \in H_\Delta\) also interpolates \(f\) as above, then \(e(x) = Q_N f(x) - h(x)\) is a cubic polynomial on \([x_i, x_{i+1}], 0 \leq i \leq N,\) and \(e(\sigma_j) = 0, 1 \leq i \leq 2N + 2.\) We show that \(e(x)\) is identically zero in \([x_i, x_{i+1}].\)

If this is not so, then without loss of generality we may assume that \(e(x) \neq 0\) for all \(x \in [x_1, x_2].\) Rolle’s Theorem implies that \(e(x_2) D_x e(x_2) > 0.\) Similarly, \(D_x e\) restricted in \([x_2, x_3]\) has roots in \((x_2, \sigma_4), (\sigma_4, \sigma_5).\) Thus, \(e(x_3) D_x e(x_3) > 0.\) By induction \(e(x_{N+1}) D_x e(x_{N+1}) > 0\) contradicting the relation \(e(x_{N+1}) = 0.\) This proves that \(e(x) \equiv 0\) in \(I.\)

2. Two-Dimensional Interpolation Scheme. In this section we introduce a two-dimensional analogue of the interpolation scheme of the previous section. Let \(\Delta_x = (y_j)_{j=0}^{M+1}\) be a partition of \([c, d], J = [c, d], k_j = [y_{j+1} - y_j], J_j = [y_{j+1}, y_{j+2}]\) and \(j = \max k_j.\) Also, we denote by \(\rho = \Delta x \Delta y\) a partition of \([a, b] \times [c, d]\) and by \(H_\rho\) the vector space of all piecewise bicubic polynomials \(p(x, y)\) with respect to \(\rho,\) such that \(D_x^2 D_y p(x, y)\) is continuous on \([a, b] \times [c, d]\) for all \(0 \leq l, \eta \leq 1.\)

The Gaussian points in the subinterval \([y_j, y_{j+1}]\) are

\[\tau_{2i+j} = \frac{y_{i+1} + y_{i+1}}{2} + \frac{(-1)^j k_i}{2}, \quad j = 1, 2.\]

A two-dimensional interpolation operator is defined as the tensor product

\[Q_\rho = Q_N \otimes Q_M = Q_N Q_M.\]

3. Error Analysis. In this section, we establish a priori bounds for the interpolation scheme introduced in Section 2 for a uniform partition \(\Delta\) of \([0, 1]\) with mesh length \(h = N^{-1}.\)

Let

\[
\phi(x) = \begin{cases} 
(1-x)^2(1+2x), & 0 \leq x \leq 1, \\
(1+x)^2(1-2x), & -1 \leq x \leq 0, \\
0, & \text{otherwise,}
\end{cases}
\]

and

\[
\psi(x) = \begin{cases} 
x(1-x)^2, & 0 \leq x \leq 1, \\
x(1+x)^2, & -1 \leq x \leq 0, \\
0, & \text{otherwise.}
\end{cases}
\]

The basis functions of the \(C^1\) cubic piecewise polynomial space \(H_\Delta\) are defined by

\[
B_{2i-1}(x) = \phi \left( \frac{x-x_i}{h} \right), \quad 1 \leq i \leq N + 1,
\]

\[
B_{2i}(x) = h \psi \left( \frac{x-x_i}{h} \right), \quad 1 \leq i \leq N + 1.
\]
For later use, we define the \textit{Gramian} matrix
\[ G_N \equiv (B_i(o_j); i, j = 1, \ldots, 2N + 2) \]
of the interpolation operator \( Q_N \). Using the \((2N + 2) \times (2N + 2)\) matrix
\[ H_N \equiv \begin{bmatrix} 1 & h & 0 \\ & & \\ & & 1 \\ 0 & & h \end{bmatrix}, \]
we find that*
\[ H_N^{-1}G_N \equiv \begin{bmatrix} 1 & A & & & \\ & & & & \\ 0 & B & & & 0 \\ & & & & \\ & & & & \\ & & & & \\ 0 & & & & \\ & B & & & A \\ & & B & & 1 \\ & & & & 0 \end{bmatrix}, \]
where
\[ A \equiv \begin{bmatrix} \alpha \\ \gamma \end{bmatrix}, \quad B \equiv \begin{bmatrix} \beta & \alpha \\ -\delta & -\gamma \end{bmatrix} \]
and
\[ \alpha = \frac{9 + 4\sqrt{3}}{18}, \quad \beta = \frac{9 - 4\sqrt{3}}{18}, \quad \gamma = \frac{3 + \sqrt{3}}{36}, \quad \delta = \frac{3 - \sqrt{3}}{36}. \]

We will also use the matrix
\[ T \equiv BA^{-1} = \begin{bmatrix} -7 & 48 \\ 1 & -7 \end{bmatrix}. \]

It is easy to see that for all integers \( n \), \( T^n \equiv I \),
\[ T^n = \begin{bmatrix} a_n & 48c_n \\ c_n & a_n \end{bmatrix}, \]
where
More generally, from \( T^{s+t} = T^s T^t \) we get

\[
\begin{align*}
    a_{s \pm t} &= a_s a_t \pm 48 c_s c_t, \\
    c_{s \pm t} &= c_s a_t \pm a_s c_t, \\
    a_{-t} &= a_p, \\
    c_{-t} &= -c_t.
\end{align*}
\]

(3.1)

Let \( \lambda_n \equiv |a_n|/c_n| = -a_n/c_n \). Since \( \det(T^n) = 1 \), we can easily show that \( \lambda_n \) is decreasing with \( n \) and for all \( n \)

\[
\sqrt{48} < \lambda_n < 7, \quad \lambda_1 = 7,
\]

(3.2)

\[
|a_{n+1}| > |a_n|, \quad |c_{n+1}| > |c_n|.
\]

Since

\[
|a_n| = \frac{1}{2}(|c_{n+1}| - |c_{n-1}|), \quad |c_n| = \frac{1}{96} (|a_{n+1}| - |a_n|),
\]

we also have

\[
\sum_{l=q}^{p} |a_l| = \frac{1}{2} (|c_{p+1}| + |c_p| - |c_q| - |c_{q-1}|),
\]

(3.3)

\[
\sum_{l=q}^{p} |c_l| = \frac{1}{96} (|a_{p+1}| + |a_p| - |a_q| - |a_{q-1}|).
\]

We introduce a \((2N + 2) \times (2N + 2)\) matrix \( R \) in partition form

\[
\begin{bmatrix}
    r_{11} & r_{12} & \cdots & r_{1,2N+1} & r_{1,2N+2} \\
    \vdots & & \ddots & \vdots & \vdots \\
    r_{N,1} & & \cdots & R_{1,N+1} \\
    r_{2N+2,1} & r_{2N+2,2} & \cdots & r_{2N+2,2N+1} & r_{2N+2,2N+2}
\end{bmatrix},
\]
where the first and last rows are defined as

\[
[r_{1,2j-1}, r_{1,2j}] = \frac{(-1)^{j+1}}{c_N} [c_{N-j+1} a_{N-j+1}],
\]

\[
[r_{2N+2,2j-1}, r_{2N+2,2j}] = \frac{(-1)^{j-N}}{c_N} [-c_{j-1} a_{j-1}],
\]

while the \(2 \times 2\) matrices \(R_{n,m}\) are defined as

\[
R_{n,m} \equiv A^{-1} \left[(-T)^{n-1} Z_m + a_{n,m} (-T)^{n-m}\right], \quad n = 1, \ldots, N, m = 1, \ldots, N + 1,
\]

with

\[
Z_1 = \begin{bmatrix} 0 & \lambda_N \\ 0 & 1 \end{bmatrix}, \quad Z_m \equiv \frac{(-1)^m}{c_N} \begin{bmatrix} c_{N-m+1} & a_{N-m+1} \\ 0 & 0 \end{bmatrix}, \quad m = 2, \ldots, N + 1,
\]

and

\[
\sigma_{n,m} = \begin{cases} 1 & \text{if } 2 \leq m \leq n, \\ 0 & \text{otherwise}. \end{cases}
\]

The reason for introducing the matrix \(R\) becomes apparent by the following lemma, which shows that we have explicitly constructed the inverse of the matrix \(H_N^{-1}G_N\).

**Lemma 3.1.** The matrix \(H_N^{-1}G_N\) is invertible and its inverse is the matrix \(R\).

**Proof.** Let \(S = R(H_N^{-1}G_N)\). It is enough to show that \(S = I\). We partition \(S\) into blocks:

\[
S = \begin{bmatrix}
s_{11} & \tau_{11} & \cdots & \tau_{1N} & s_{1,2N+2} \\
\omega_{11} & S_{11} & \cdots & S_{1N} & \omega_{1,2N+2} \\
. & . & \cdots & . & . \\
. & . & \cdots & . & . \\
\omega_{N1} & S_{N1} & \cdots & S_{NN} & \omega_{N,2N+2} \\
s_{2N+2,1} & \tau_{2N+2,1} & \cdots & \tau_{2N+2,N} & s_{2N+2,2N+2}
\end{bmatrix},
\]

where each \(s_{ij}\) is \(1 \times 1\), \(S_{ij}\) is \(2 \times 2\), \(\omega_{ij}\) is \(2 \times 1\) and \(\tau_{ij}\) is \(1 \times 2\). Performing the multiplication of the matrices \(R\) and \(H_N^{-1}G_N\), we obtain

\[
s_{11} = r_{11} = 1.
\]
\[ \tau_{ij} = [s_{1,2j} s_{1,2j+1}] = [r_{1,2j-1} r_{1,2j}] A + [r_{1,2j-1} r_{1,2j+2}] B \]

\[ = \frac{(-1)^j}{c_N} \{ [c_{N-j+1} a_{N-j+1}] - [c_{N-j} a_{N-j}] T \} A \]

\[ = \frac{(-1)^j}{c_N} \{ [c_{N-j+1} a_{N-j+1}] - [c_{N-j+1} a_{N-j+1}] \} A = [0, 0], \]

and

\[ s_{1,2N+2} = r_{1,2N+1} = 0. \]

Similarly

\[ \omega_{i,1} = \omega_{i,2N+2} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \quad \tau_{2N+2,j} = [0 0], \quad i, j = 1, \ldots, N, \]

and

\[ s_{2N+2,1} = 0, \quad s_{2N+2,2N+2} = 1. \]

For the square blocks \( S_{n,m} \) we find

\[ S_{n,m} = R_{n,m} A + R_{n,m+1} B \]

\[ = A^{-1} (-T)^{n-1} (Z_m + Z_{m+1} T + (a_{n,m} - a_{n,m+1}) (-T)^{1-m}) A. \]

From the definition of \( Z_m \) and \( T \) we obtain \( Z_m + Z_{m+1} T = \delta_m I \). Then from the definition of \( a_{n,m} \) we get \( S_{n,m} = \delta_m I \). This concludes the proof of Lemma 3.1.

**Lemma 3.2.** If \( G_N \) is the Gramian of the interpolation operator \( \mathcal{Q}_N \), then

\[ \| (H_N^{-1} G_N)^{-1} \|_m < 100 \]

for all \( N \geq 2 \).

**Proof.** Let

\[ \| R \|_1 = \sum_{m=1}^{2N+2} |r_{lm}|. \]

From the definition of \( R \) and relations (3.1), (3.2), (3.3), we obtain

\[ \| R \|_1 = \sum_{j=1}^{N} (|r_{1,2j+1}| + |r_{1,2j}|) \]

\[ = \frac{1}{|c_N|} \sum_{j=1}^{N} (|c_{N-j+1}| + |a_{N-j+1}|) = \frac{1}{|c_N|} \sum_{i=1}^{N} (|c_i| + |a_i|) \]

\[ \leq \frac{7}{12} \frac{|a_N|}{|c_N|} + \frac{9}{2} + \frac{5}{15} \frac{|a_1|}{|c_N|} - \frac{7}{2} \frac{|c_1|}{|c_N|} \leq 23/2. \]

It is easy to see that \( \| R \|_{2N+2} = \| R \|_1 \). For the remaining rows we use (3.1), (3.2) to get that for \( 2 \leq m \leq n \),
\( AR_{n,m} = (-T)^{n-1} Z_m + \sigma_{n,m} (-T)^{n-m} \)

\[
= \frac{1}{2 \| e_N \|} \begin{bmatrix}
-|c_{N-n-m+2}| + |c_{N-n+m}| & |a_{N-n-m+2}| + |a_{N-n+m}| \\
\frac{1}{48} (-|a_{N-n-m+2}| + |a_{N-n+m}|) & |c_{N-n-m+2}| + |c_{N-n+m}|
\end{bmatrix},
\]

while for \( n < m \)

\[
AR_{n,m} = \frac{1}{2 \| e_N \|} \begin{bmatrix}
-|c_{N+n-m}| + |c_{N-n-m+2}| & |a_{N+n-m}| + |a_{N-n-m+2}| \\
\frac{1}{48} (|a_{N+n-m}| - |a_{N-n-m+2}|) & |c_{N+n-m}| - |c_{N-n-m+2}|
\end{bmatrix}.
\]

Finally, for \( m = 1 \),

\[
AR_{n,1} = \frac{1}{\| e_N \|} \begin{bmatrix}
0 & |a_{N-n+1}| \\
0 & |c_{N-n+1}|
\end{bmatrix}.
\]

Using again the relations (3.1) through (3.3), we now find

\[
\sum_{m=1}^{N} \| AR_{n,m} \|_1 \leq \frac{1}{2} \left[ 2 \frac{|a_{N-n+1}|}{\| e_N \|} + \frac{1}{96} \left( \frac{|a_{N-n+1}|}{\| e_N \|} + \frac{|a_{N-n}|}{\| e_N \|} + \frac{|a_{N-1}|}{\| e_N \|} \right) \right] + \frac{1}{2} \left( \frac{|c_{N-n}|}{\| e_N \|} + 9 + \frac{|c_{N-1}|}{\| e_N \|} \right) \leq \frac{35}{3}
\]

and

\[
\sum_{m=1}^{N} \| AR_{n,m} \|_2 \leq \frac{1}{2} \left[ 2 \frac{|c_{N-n+1}|}{\| e_N \|} + \frac{1}{96} \left( \frac{|a_{N-n}|}{\| e_N \|} + \frac{|a_{N-1}|}{\| e_N \|} + 48 + \frac{|c_{N-n+1}|}{\| e_N \|} \right) \right] + \frac{1}{2} \left( \frac{|c_{N-n}|}{\| e_N \|} + \frac{|a_{N-1}|}{\| e_N \|} + \frac{|a_{N}|}{\| e_N \|} + \frac{|c_{N-1}|}{\| e_N \|} \right) \leq 2.
\]

By definition now, we have for \( l = 1, 2 \),

\[
\| R \|_{2^n+1} = \sum_{m=1}^{N} \| A^{-1} AR_{n,m} \|_1 \leq \sum_{m=1}^{N} \| A^{-1} \|_\infty \| AR_{n,m} \|_1,
\]

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while \( \| A^{-1} \|_\infty = (7\sqrt{3} + 9)/4 \). Thus, for the norm \( \| R \|_\infty = \max_i \| R_i \| \) the following bound holds

\[
\| R \|_\infty = \|(H^{-1}G_N)^{-1}\|_\infty < 100.
\]

This concludes the proof of Lemma 3.2.

*Remark.* As the proof of Lemma 3.2 suggests, the bound (3.4) can be improved. Our conjecture is that a more careful analysis will show that the norm \( \|(H_N^{-1}G_N)^{-1}\|_\infty \) is decreasing in \( N \), that

\[
\lim_{N \to \infty} \|(H_N^{-1}G_N)^{-1}\|_\infty = \frac{69 - 29\sqrt{3}}{2}
\]

and that for all \( N \geq 2 \)

\[
\frac{69 - 29\sqrt{3}}{2} \leq \|(H_N^{-1}G_N)^{-1}\|_\infty \leq \|(H_2^{-1}G_2)\|_\infty = \frac{33\sqrt{3} + 9}{7}.
\]

Numerical experiments confirm this conjecture.

**Lemma 3.3.** Let \( Q_N \) be the interpolation operator defined by (1.2). Then \( Q_N \) is bounded in the \( L_\infty \)-norm.

*Proof.* We can express the \( Q_Nf \) interpolant as

\[
(Q_Nf)(x) = \sum_{i=1}^{N+1} \{a_iB_{2i-1}(x) + b_iB_{2i}(x)/h\},
\]

where the \( B_i \)’s are the previously defined basis functions of \( H_\Delta \). Let \( f \) denote the column vector \( \{f(a_j)\}_{j=1}^{N+2} \) and set \( L_N = (H_N^{-1}G_N)^T \). Then,

\[
f = L_N \begin{bmatrix} a_1 \\ b_1 \\ a_2 \\ b_2 \\ a_3 \\ b_3 \\ \vdots \end{bmatrix}.
\]

In Lemma 3.2 we prove that \( \|L_N^{-1}\|_\infty < 100 \); therefore,

\[
\max_i (\max(|a_i|, |b_i|)) \leq 100 \|f\|_{L_\infty}.
\]

For all \( x \in [x_i, x_{i+1}] \), \( 1 \leq i \leq N \), we have

\[
(Q_Nf)(x) = a_iB_{2i-1}(x) + b_iB_{2i}(x)/h + a_{i+1}B_{2i+1}(x) + b_{i+1}B_{2i+2}(x)/h;
\]

and since \( B_{2i-1}, B_{2i}/h \) are bounded by unity, we obtain
Theorem 3.1. If \( f \in W^{s,\infty}(I) \), \( s = 0, 1, 2, 3, \) or \( 4 \), then

(i) \( Q_Nf \to f \), as \( N \to \infty \), and

(ii) for the interpolation error we have

\[
\|Q_Nf - f\|_{L_\infty} \leq c\|D^s f\|_{L_\infty},
\]

where \( c \) is independent of \( h \).

Proof. Let \( \partial_H f \) be the Hermite interpolant of \( f \), defined by interpolation of \( f \) and its first derivative at the nodes of the partition \( \Delta \). From the triangle inequality we find

\[
\|f - Q_Nf\|_{L_\infty} \leq (1 + \|Q_N\|)\|f - \partial_H f\|_{L_\infty}.
\]

Moreover, for the Hermite interpolation error, it is known [3, p. 236]

\[
\|f - \partial_H f\|_{L_\infty} \leq c h^s\|D^s f\|_{L_\infty}, \quad s > 0.
\]

From (3.5), (3.6), (3.7) and Lemma 3.3, we now get

\[
\|f - Q_Nf\|_{L_\infty} \leq c h^s\|D^s f\|_{L_\infty}.
\]

This proves conclusions (i) and (ii).

Theorem 3.2. If \( f \in W^{s,\infty}(\Omega) \), \( s = 1, 2, 3, 4 \), and \( \Omega = [0, 1]^2 \), then

\[
\|f - Q_P f\|_{L_\infty} \leq c(h^p\|D^p f\|_{L_\infty} + h^q k^q\|D^q D^p f\|_{L_\infty} + k^s\|D^s f\|_{L_\infty})
\]

(3.8)

\[
\leq c p^s\|f\|_{W^{s,\infty}}, \quad \text{where} \quad s = p + q
\]

and \( p, q \) are integers.

Proof. From the triangle inequality we have

\[
\|f - Q_P f\|_{L_\infty} \leq \|f - Q_N f\|_{L_\infty} + \|Q_N(f - Q_M f)\|_{L_\infty}
\]

\[
\leq \|f - Q_N f\|_{L_\infty} + \|Q_N(f - Q_M f)\|_{L_\infty} + \|f - Q_M f\|_{L_\infty}.
\]

Using the results of Theorem 3.1, we now get

\[
\|f - Q_P f\|_{L_\infty} \leq h^p\|D^p f\|_{L_\infty} + h^q k^q\|D^q D^p f\|_{L_\infty} + k^s\|D^s f\|_{L_\infty}
\]

\[
\leq h^p\|D^p f\|_{L_\infty} + h^q k^q\|D^q D^p f\|_{L_\infty} + k^s\|D^s f\|_{L_\infty}
\]

\[
\leq c p^s\|f\|_{W^{s,\infty}},
\]

which completes the proof of the theorem.
II. Two-Dimensional Collocation

4. Procedure and Error Estimation. In this section we consider the problem of approximating the solution of the integral equation

\[ Au = u(P) - \lambda \int_{\Omega} k(P; Q) u(Q) dQ = f(P) \quad \text{where } \Omega = [0, 1]^2 \]

and, for brevity, \( P = (x, y) \), \( Q = (s, t) \) and \( dQ = ds \, dt \).

Let

\[ Ku \equiv \int_{\Omega} k(P; Q) u(Q) dQ. \]

We seek an approximation \( u_{\rho} \in H_{\rho} \) to \( u \) of the form

\[ u_{\rho}(P) = \sum_{i=1}^{2N+2} \sum_{j=1}^{2M+2} \alpha_{ij} B_i(x) B_j(y) \]

such that

\[ (I - \rho Q_{\rho} K) u_{\rho} = Q_{\rho} f. \]

**Theorem 4.1.** If

A1. \( \lambda \) is not an eigenvalue of the kernel \( k(P; Q) \).
A2. The right side and the kernel of Eq. (4.1) are in \( W^{s, \infty}(\Omega) \), \( s = 1, 2, 3, \) or 4,

then

(i) for sufficiently small \( |\rho| \) the collocation system (4.3) is uniquely solvable and
(ii) for the error of approximation we have

\[ \|u - u_{\rho}\|_{L_\infty} \leq c\|u - Q_{\rho} u\|_{L_\infty} \leq c\rho^3 \|u\|_{W^{s, \infty}}. \]

**Proof.** By the definition of the operator \( K \) we obtain

\[ \|Ku - Q_{\rho} Ku\|_{L_\infty} = \left\| \int_{\Omega} (k(\cdot, Q) - Q_{\rho} k(\cdot, Q)) u dQ \right\|_{L_\infty} \]

and since \( \|k(\cdot, Q) - Q_{\rho} k(\cdot, Q)\|_{L_\infty} \to 0 \) as \( |\rho| \to 0 \) it follows from (4.5) that

\[ \|K - Q_{\rho} K\| \to 0 \quad \text{as} \quad |\rho| \to 0. \]

In fact, for any \( \epsilon > 0 \) there exists \( \delta_0(\epsilon) \) such that for \( |\rho| < \delta_0(\epsilon) \) the inequality \( \|K - Q_{\rho} K\| < \epsilon \) holds. But in this case, on the sphere \( \|u\|_{L_\infty} = 1 \) we have

\[ \|(I - \lambda Q_{\rho} K) u\|_{L_\infty} \geq \|(I - \lambda K) u\|_{L_\infty} - |\lambda| \|(K - Q_{\rho} K) u\|_{L_\infty} \geq \alpha - |\lambda| \epsilon, \]

where \( 0 < \alpha \leq \inf_{\|u\|_{L_\infty} = 1} \|(I - \lambda K) u\|_{L_\infty}. \)

Consequently, for sufficiently small \( |\rho| \), the relationship

\[ \inf_{\|u\|_{L_\infty} = 1} \|(I - \lambda Q_{\rho} K) u\|_{L_\infty} \geq \beta > 0 \]

holds, from which conclusion (i) follows.
Now from (4.3) it follows easily that

\[ u - u_\rho = (I - \lambda Q_\rho)^{-1}(u - Q_\rho u) \]

from which

\[ \|u - u_\rho\|_{L^\infty} \leq (I - \lambda Q_\rho)^{-1}\|u - Q_\rho u\|_{L^\infty} \leq \frac{1}{\beta} \|u - Q_\rho u\|_{L^\infty} \]

and combining this inequality with that of (3.8) we obtain (4.4). This concludes the proof of Theorem 4.1.

Finally, we remark that the above results also hold for one-dimensional integral equations of the second kind.

5. Numerical Results. In this section we present some numerical results concerning the approximation of the solution of some one-dimensional integral equations taken from [2]. The numerical solutions are computed by one-dimensional analogue of the collocation scheme introduced in Part II with a three-point Gaussian rule and by Nyström’s method with Simpson’s numerical integration rule [1]. The partition \( \Delta \) used is uniform with mesh size \( h = 1/N \). The rate of convergence estimate

\[ \log \left( \frac{\text{error for } h}{\text{error for } h/2} \right) / \log 2 \]

is also given.

The integral equation

\[ u(s) - \lambda \int_a^b k(s, t)u(t)dt = f(s), \quad a \leq s \leq b, \]

is solved for various kernel functions \( k \), right side functions \( f \) and parameters \( \lambda \).

Case (i). \( k(s, t) = \cos(\pi st), 0 < s, t < 1, \lambda = 1 \). The right side \( f \) is chosen so that \( u(s) = e^s \cos(\pi s) \).

<table>
<thead>
<tr>
<th>( N )</th>
<th>Max. Error</th>
<th>Rate</th>
<th>Max. Error</th>
<th>Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>3</td>
<td>4.55 \times 10^-2</td>
<td></td>
<td>1.18 \times 10^-2</td>
<td></td>
</tr>
<tr>
<td>6</td>
<td>4.24 \times 10^-3</td>
<td>3.4</td>
<td>5.85 \times 10^-4</td>
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<td>3.37 \times 10^-4</td>
<td>3.7</td>
<td>3.47 \times 10^-5</td>
<td>4.1</td>
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<tr>
<td>24</td>
<td>2.59 \times 10^-5</td>
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<td>2.14 \times 10^-6</td>
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</tr>
<tr>
<td>48</td>
<td>1.75 \times 10^-6</td>
<td>3.9</td>
<td>1.33 \times 10^-7</td>
<td>4.</td>
</tr>
</tbody>
</table>

Case (ii). \( k(s, t) = e^{\beta st}, 0 \leq s, t \leq 1, \lambda = 1 \). For the numerical example we pick \( f \) so that

\[ u(s) = e^{\alpha s}, \quad \alpha = 1, \beta = 5. \]
<table>
<thead>
<tr>
<th>N</th>
<th>Max. Error</th>
<th>Rate</th>
<th>Max. Error</th>
<th>Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>3</td>
<td>1.33 x 10^{-3}</td>
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<td>1.60 x 10^{-1}</td>
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<tr>
<td>6</td>
<td>2.17 x 10^{-5}</td>
<td>5.9</td>
<td>1.55 x 10^{-2}</td>
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<td>1.02 x 10^{-3}</td>
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<td>1.1 x 10^{-8}</td>
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<td>6.49 x 10^{-5}</td>
<td>3.97</td>
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<tr>
<td>48</td>
<td>8.57 x 10^{-10}</td>
<td>3.7</td>
<td>4.06 x 10^{-6}</td>
<td>4.00</td>
</tr>
</tbody>
</table>

Case (iii). \( k(s, t) = t - s, \ 0 \leq t, s \leq 1, \lambda = 1 \). Choose \( f \) so that

\[
u(s) = s^{\alpha/2}, \quad \alpha = 1, 3, 5, 7, 9.\]

<table>
<thead>
<tr>
<th>( \alpha = 1 )</th>
<th>COLLOCATION</th>
<th>NYSTROM</th>
</tr>
</thead>
<tbody>
<tr>
<td>N</td>
<td>Max. Error</td>
<td>Rate</td>
</tr>
<tr>
<td>3</td>
<td>9.79 x 10^{-3}</td>
<td>.5</td>
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<tr>
<td>6</td>
<td>6.92 x 10^{-3}</td>
<td>.5</td>
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<tr>
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<td>4.87 x 10^{-3}</td>
<td>.5</td>
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<tr>
<td>24</td>
<td>3.43 x 10^{-3}</td>
<td>.5</td>
</tr>
<tr>
<td>48</td>
<td>2.42 x 10^{-3}</td>
<td>.5</td>
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</tbody>
</table>

<table>
<thead>
<tr>
<th>( \alpha = 3 )</th>
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<th>NYSTROM</th>
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</thead>
<tbody>
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<td>Max. Error</td>
<td>Rate</td>
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<tr>
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<td>1.46 x 10^{-4}</td>
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<tr>
<td>12</td>
<td>5.10 x 10^{-5}</td>
<td>1.5</td>
</tr>
<tr>
<td>24</td>
<td>1.79 x 10^{-5}</td>
<td>1.5</td>
</tr>
<tr>
<td>48</td>
<td>6.32 x 10^{-6}</td>
<td>1.5</td>
</tr>
</tbody>
</table>
Case (iv).

\[ \lambda = 0.3 \text{ and } k(s, t) = \begin{cases} -s(1 - t), & 0 \leq s \leq t < 1, \\ -t(1 - s), & 0 \leq t \leq s < 1. \end{cases} \]
Choose \( f \) so that

\[
u(s) = 25s^5(1 - s).
\]

<table>
<thead>
<tr>
<th>N</th>
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<th>NYSTROM</th>
</tr>
</thead>
<tbody>
<tr>
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<td>Rate</td>
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<tr>
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</tr>
<tr>
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<td>(4.75 \times 10^{-2})</td>
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<td>6</td>
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<td>3.6</td>
</tr>
<tr>
<td>48</td>
<td>(2.22 \times 10^{-6})</td>
<td>3.5</td>
</tr>
</tbody>
</table>

Case (v). \( k(s, t) = t - s, \ 0 \leq s, t \leq 1, \lambda = 1. \)

Choose \( f \) so that

\[
u(s) = s \ln(s)
\]

<table>
<thead>
<tr>
<th>N</th>
<th>COLLOCATION</th>
<th>NYSTROM</th>
</tr>
</thead>
<tbody>
<tr>
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<td>----</td>
<td>-------------------</td>
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<tr>
<td>24</td>
<td>(1.86 \times 10^{-3})</td>
<td>1.0</td>
</tr>
<tr>
<td>48</td>
<td>(9.09 \times 10^{-4})</td>
<td>1.0</td>
</tr>
</tbody>
</table>

The estimated rates of convergence in Cases (i)–(v) are in good agreement with those suggested by Theorem 4.1 depending on the smoothness of the solution.

The above data indicate that the collocation method is faster than Nyström for problems with smooth solutions and nonsmooth kernels. The Nyström method runs faster in the cases of nonsmooth solutions but we believe that collocation with non-uniform mesh will do equally well. It is worth noting that for both methods all the time is spent in solving the linear system.

The Cases (i)–(v) were solved by the collocation method described in this paper with a two-point Gaussian rule. The results obtained are less accurate, but the computed rate estimates agree with the a priori estimate obtained in Theorem 4.1 as expected.
All the numerical experiments were carried out on a CDC 6500 in single precision.

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12. E. N. HOUSTIS & T. S. PAPATHODOPOULOU, Piecewise Cubic Hermite Interpolation at the Gaussian Points, Purdue University, CSD-TR 199, July 1976.