

FINITE DIFFERENCE PRECONDITIONING FOR SOLVING ORTHOGONAL COLLOCATION EQUATIONS FOR BOUNDARY VALUE PROBLEMS*WEIWEI SUN[†], WEIZHANG HUANG[‡], AND ROBERT D. RUSSELL[§]

Abstract. A technique to construct a low-order finite difference preconditioner for solving orthogonal collocation equations for boundary value problems is presented. It is shown numerically and theoretically that the spectral condition numbers of the preconditioned collocation matrices are bounded by constants independent of the number of mesh nodes when certain exact low-order finite difference preconditionings are used. Preconditioners based on incomplete LU factorization are also discussed. Numerical experiments show the efficiency and robustness of the preconditioning.

Key words. collocation, preconditioning, boundary value problem

AMS subject classifications. 65N22, 65N35, 65L60

1. Introduction. In this paper, we shall consider orthogonal collocation methods for solving elliptic differential equations. These methods have been extensively investigated in the last 20 years due to their ease of implementation and high-order accuracy (see [RS72], [DS73], and [PR76]). An apparent disadvantage of orthogonal collocation lies in the fact that a collocation matrix is generally nonsymmetric, even for a Poisson equation with a uniform mesh. Gauss elimination has often been used for solving the collocation equations, and its cost has restricted the applicability of the methods.

In recent years, some efforts in solving the collocation equations by alternative methods have been made. A so-called alternating direction collocation (ADC) or generalized implant (ADI) method was introduced by Hayes [Hay80] and Hayes, Pinder, and Celia [HPC81] for solving a parabolic problem. A full theory for convergence of ADC on a uniform mesh is given for the model problem $\Delta u = f$ in a rectangular domain by Dyksen [Dyk87]. Significantly, a closed form for the eigenvalues and their associated eigenfunctions is also given. Cooper and Prenter [CP91] give a complete theory for convergence of ADC applied to a large class of linear separable elliptic partial differential equations (PDEs) on a rectangular domain. Bialecki [Bia91] considers the problem of determining optimal acceleration parameters for ADC by using the Jordan acceleration technique. Besides ADC, a fast direct method (FFT) has been developed for solving $\Delta u = f$ on a uniform mesh in [SZ89] and [BFB92]. In [Sun94a], a class of block stationary (standard) methods are applied to second-order separable elliptic PDEs and their convergence is proven. However, all of the above methods are only applicable for linear separable and self-adjoint elliptic PDEs.

It has been well demonstrated that when solving PDEs a significant improvement in convergence of conventional iterative methods can be gained by using preconditioning techniques. They have been investigated extensively for finite difference, finite element, and spectral methods; e.g., see [Ors80], [Eva83], [Won87], [CE88], and [CHQZ88]. For spectral methods, a low-order finite difference approximation or its incomplete LU factorization can

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often be proven to be an efficient preconditioning technique [Ors80]. Recently, for finite element and finite difference methods, some theoretical analysis applicable to a large class of elliptic PDEs has been carried out in [MP90] and [JMPW92]. However, to our knowledge preconditioning has not been applied in the orthogonal collocation context. The objective of this paper is to present a technique to construct such a finite difference preconditioning. A major difficulty lies in the fact that the unknowns in the Hermite collocation approximation involve both the variable u and its derivatives, while those in the standard finite difference approximation contain only the variable u .

An outline of the paper is as follows. In §2, we introduce some notation and give a brief description of the orthogonal Hermite cubic collocation method. In §3, a technique to construct a finite difference preconditioning is presented for one- and two-dimensional (1D and 2D) model problems. Several versions of the preconditioning are discussed. In §4, we show that for the model problems with periodic boundary conditions and a uniform mesh, the condition number of a preconditioned system is bounded by a constant independent of the number of mesh nodes. Numerical results supporting this theory and showing the efficacy of the preconditioning methods are presented in §5. Section 6 contains conclusions and additional comments.

2. Orthogonal collocation methods. Consider the problem

$$(2.1) \quad Lu := -u'' + \alpha u = f(x), \quad x \in (0, 1)$$

with the Dirichlet boundary conditions $u(0) = u(1) = 0$ or the periodic boundary conditions $u(0) = u(1)$ and $u'(0) = u'(1)$. Given a partition $\Pi_N = \{x_i\}_{i=0}^N$ on $[0,1]$ such that

$$0 = x_0 < x_1 < \dots < x_N = 1,$$

the Hermite cubic collocation approximation v is defined as follows:

$$(2.2) \quad v(x) = \xi_1(s)u_i + h_i\xi_2(s)u'_i + \xi_1(1-s)u_{i+1} - h_i\xi_2(1-s)u'_{i+1}, \quad s \in [0, 1]$$

on each element $[x_i, x_{i+1}]$ ($i = 0, \dots, N - 1$), where $s = (x - x_i)/h_i$, $h_i = x_{i+1} - x_i$, u_i and u'_i denote approximations to $u(x_i)$ and $u'(x_i)$, respectively, and $\xi_1(s)$ and $\xi_2(s)$ are the so-called shape functions for 1D Hermite cubic interpolation defined by

$$\begin{aligned} \xi_1(s) &= (1 + 2s)(1 - s)^2, \\ \xi_2(s) &= s(1 - s)^2. \end{aligned}$$

Denoting the collocation points by $\Pi_{COL} := \{x_j^*\}_{j=0}^{2N+1}$, where

$$(2.3) \quad \begin{aligned} x_0^* &= x_0 - \sigma(x_1 - x_0), \\ x_{2i+1}^* &= x_i + \sigma(x_{i+1} - x_i), & i = 0, \dots, N - 1, \\ x_{2i+2}^* &= x_i + (1 - \sigma)(x_{i+1} - x_i), & i = 0, \dots, N - 1, \\ x_{2N+1}^* &= x_N + \sigma(x_N - x_{N-1}), \end{aligned}$$

with $\sigma := \frac{1}{2}(1 - \frac{1}{\sqrt{3}})$ and $1 - \sigma$ (the Gaussian points on $(0,1)$), the orthogonal Hermite cubic collocation approximation $v(x)$ of (2.1) is defined by the collocation equations

$$(2.4) \quad Lv(x_{2i+l}^*) = -v''(x_{2i+l}^*) + \alpha v(x_{2i+l}^*) = f(x_{2i+l}^*), \quad l = 1, 2; \quad i = 0, \dots, N - 1,$$

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and boundary conditions $v(x_0) = v(x_{2N+1}) = 0$. This can be rewritten in the matrix form

$$(2.5) \quad A_{COL}v := A_C v + \alpha B_C v = f,$$

where

$$(2.6) \quad \begin{aligned} v &= [u'_0, u_1, u'_1, \dots, u_{N-1}, u'_{N-1}, u'_N]^T, \\ f &= [f(x_1^*), \dots, f(x_{2N}^*)]^T. \end{aligned}$$

The matrices A_C and B_C can be regarded as collocation approximations to the second derivative and identity operators, respectively, and are neither symmetric nor diagonally dominant. This is why, although some iterative and fast FFT-like methods have been developed recently for solving many PDE systems, Gauss elimination has predominated in the general collocation matrix context.

3. Finite difference preconditioning. In this section, we present a technique to construct a finite difference preconditioning for the orthogonal collocation system. For simplicity, the technique is illustrated for orthogonal Hermite cubic collocation applied to Dirichlet and periodic problems for 1D and 2D Poisson equations. However, as will be evident, the technique may be applied for more general collocation methods and boundary value problems.

3.1. 1D case. We consider the 1D Dirichlet model problem (2.1). The collocation approximation contains $2(N + 1)$ unknowns, and we consider a finite difference approximation which involves the same number of unknowns. A natural way to construct such an approximation is to use centered differences based on the collocation points $\{x_j^*, j = 0, \dots, 2N + 1\}$. That is, use

$$(3.1) \quad -\frac{2}{x_{j+1}^* - x_{j-1}^*} \left[\frac{w_{j+1} - w_j}{x_{j+1}^* - x_j^*} - \frac{w_j - w_{j-1}}{x_j^* - x_{j-1}^*} \right] + \alpha w_j = f(x_j^*), \quad j = 1, 2, \dots, 2N$$

supplemented with the boundary conditions $w_0 = w_{2N+1} = 0$, where w_j denotes the finite difference approximation of $u(x_j^*)$. In matrix form this is

$$(3.2) \quad A_F w = f,$$

where f is defined by (2.6) and

$$(3.3) \quad w = [w_1, \dots, w_{2N}]^T.$$

Since w and v approximate different quantities, A_F cannot be expected to be a suitable preconditioner for A_{COL} . However, a relation between these two sets of unknowns is easily found using (2.2), viz., we have

$$(3.4) \quad \begin{aligned} v(x_{2i+1}^*) &= \xi_1(\sigma)u_i + h_i \xi_2(\sigma)u'_i + \xi_1(1 - \sigma)u_{i+1} - h_i \xi_2(1 - \sigma)u'_{i+1}, \\ v(x_{2i+2}^*) &= \xi_1(1 - \sigma)u_i + h_i \xi_2(1 - \sigma)u'_i + \xi_1(\sigma)u_{i+1} - h_i \xi_2(\sigma)u'_{i+1}, \\ & \quad i = 0, 1, \dots, N - 1. \end{aligned}$$

Both $v(x_j^*)$ and w_j are approximations to $u(x_j^*)$, and if $v(x_j^*)$ is replaced by w_j we obtain

$$(3.5) \quad w = B_C v.$$

Substituting (3.5) into (2.5) gives

$$(3.6) \quad A_{COL} B_C^{-1} w = f.$$

TABLE 1
 Condition numbers of the collocation and preconditioned matrices for the 1D model problem.

MESH NO.	N	$\kappa_1(A_{COL})$ Dirichlet	$\kappa_1(T_1)$ ($\alpha = 0$)	$\kappa_1(A_{COL})$ Periodic	$\kappa_1(T_1)$ ($\alpha = 1$)	h_1^*	h_2^*
1 uniform	8	0.265E2	3.168	0.239E3	2.196	1.0	1.0
	16	0.983E2	3.168	0.883E3	2.196	1.0	1.0
	32	0.366E3	3.168	0.335E4	2.196	1.0	1.0
2	8	0.159E4	3.123	0.743E4	0.903E1	4.0	49.0
	16	0.223E5	3.144	0.116E6	0.182E2	4.0	225.0
	32	0.326E6	3.156	0.186E7	0.364E2	4.0	961.0
3	8	0.753E7	3.743	0.310E8	0.205E4	16.0	2401.0
	16	0.164E10	3.816	0.772E10	0.188E5	16.0	50625.0
	32	0.367E12	3.854	0.171E13	0.162E6	16.0	923521.0
4 $\beta = 10$	8	0.119E4	3.743	0.156E5	3.162	8.0	10.0
	16	0.595E4	3.176	0.451E5	3.161	8.0	10.0
	32	0.265E5	3.168	0.179E6	3.161	8.0	10.0
4 $\beta = 100$	8	0.104E6	25.08	0.114E7	25.11	80.0	100.0
	16	0.600E6	25.02	0.447E7	25.11	80.0	100.0
	32	0.265E7	25.09	0.179E8	25.11	80.0	100.0
4 $\beta = 1000$	8	0.118E8	244.7	0.115E9	245.8	800.0	1000.0
	16	0.599E8	244.9	0.452E9	245.8	800.0	1000.0
	32	0.264E9	244.9	0.179E10	245.8	800.0	1000.0

FIG. 1. 1D Mesh 4 for $N = 8$.

Now (3.6) has the same unknowns as (3.2), which motivates using A_F as our preconditioner for $A_{COL}B_C^{-1}$. The periodic problem can be considered similarly.

The accuracy of this preconditioner has been examined for several examples. Throughout, the numerical experiments are performed on a Sun SPARC+1 in double precision, using standard IMSL routines for solving the linear systems and eigenvalue problems.

In Table 1, the condition numbers of A_{COL} and $T_1 := A_F^{-1}(A_{COL}B_C^{-1})$ for Dirichlet boundary conditions with $\alpha = 0$ and periodic boundary conditions with $\alpha = 1$ are listed. The condition number used here is defined as

$$\kappa_1(T) = \frac{\max_i |\lambda_i(T)|}{\min_i |\lambda_i(T)|},$$

and the meshes are the following:

- 1D Mesh 1: uniform mesh, $x_i = \frac{i}{N}, i = 0, 1, \dots, N$;
- 1D Mesh 2: $x_i = (\frac{i}{N})^2, i = 0, 1, \dots, N$;
- 1D Mesh 3: $x_i = (\frac{i}{N})^4, i = 0, 1, \dots, N$;
- 1D Mesh 4: a uniform mesh consisting of $N - 3$ nodes, with four additional local uniform nodes close to $x = 0.5$ (see Figure 1). The ratio of these two spacings is denoted by β .

The two quantities

$$(3.7) \quad \begin{aligned} h_1^* &= \max_i \left\{ \max \left\{ \frac{h_i}{h_{i+1}}, \frac{h_{i+1}}{h_i} \right\} \right\}, \\ h_2^* &= \frac{\max h_i}{\min h_i}, \end{aligned}$$

the so-called local mesh ratio, and global mesh ratio are used to measure the smoothness of a given mesh. In Figures 2 and 3, we plot the distributions of eigenvalues of A_{COL} and T_1

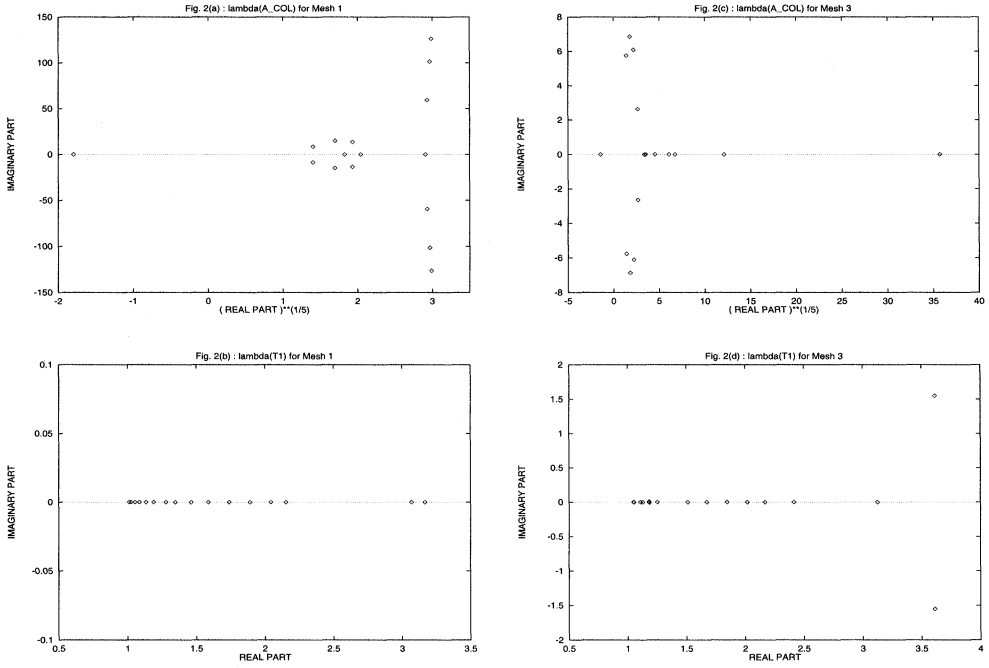


FIG. 2. Eigenvalues for the 1D model problem with Dirichlet boundary conditions ($N = 8, \alpha = 0$).

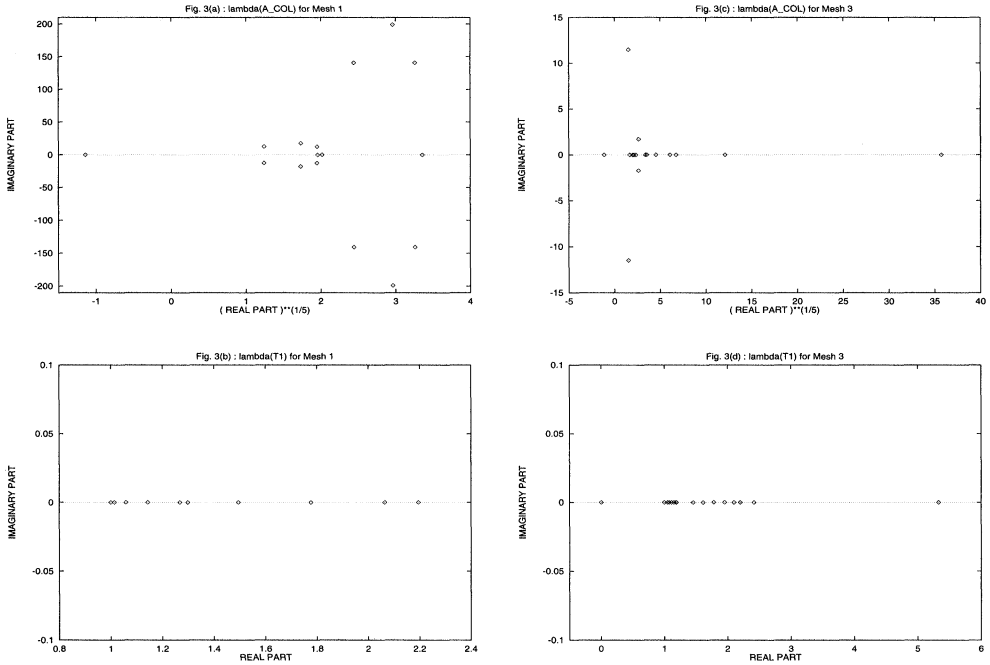


FIG. 3. Eigenvalues for the 1D model problem with periodic boundary conditions ($N = 8, \alpha = 1$).

for 1D Meshes 1 and 3 and $N = 8$ with Dirichlet and periodic boundary conditions. Several observations can be made.

Many of the eigenvalues of A_{COL} are complex. A_{COL} is indefinite, as its eigenvalues have both positive and negative real parts. This illustrates why conventional iterative methods, such as Jacobi, Gauss–Seidel and SOR, are divergent when they are applied to the collocation equations (2.5). The condition number of A_{COL} is quite sensitive to the smoothness of the mesh used.

Since both $A_{COL}B_C^{-1}$ and A_F are symmetric for 1D Mesh 1 [Bia91], the eigenvalues of T_1 are real and positive. In the case of Dirichlet boundary conditions, the condition numbers of T_1 are less sensitive than A_{COL} to the smoothness of the mesh. For a smooth or a moderately nonsmooth mesh (h_1^* is not large and the spacing does not have a steep change, such as 1D Meshes 1, 2, 3, and 4 with $\beta = 10$), $\kappa_1(T_1)$ is small. Even for a fairly nonsmooth mesh (such as 1D Mesh 4 with $\beta = 100$ or 1000, whose spacing has a steep jump near $x = 0.5$), $\kappa_1(T_1)$ is much smaller than $\kappa_1(A_{COL})$. For all meshes, the norms of the eigenvalues of the preconditioned matrix are bounded below by one and above by a constant which is independent of N . For example, the upper bound is about 3.17 for 1D Mesh 1, 3.16 for 1D Mesh 2, and 3.86 for 1D Mesh 3 with Dirichlet boundary conditions. A similar observation can be made for the periodic boundary value problem except for Mesh 2 and Mesh 3 which are not suitable to this case.

Thus A_F is a good preconditioner for $A_{COL}B_C^{-1}$. The preconditioned matrix has a desirable distribution of eigenvalues for smooth and moderately nonsmooth meshes. Some of these numerical observations will be confirmed theoretically in §4. In §5 it is demonstrated that fast convergence can be gained when a conventional iterative method, such as Richardson and conjugate gradient iteration, is applied to the preconditioned system.

3.2. 2D case. Now we illustrate the preconditioning technique for the more interesting case of a 2D Helmholtz equation

$$(3.8) \quad -\Delta u + \alpha u = f(x, y), \quad (x, y) \in \Omega$$

with Dirichlet or periodic boundary conditions on $\partial\Omega$ where $\Omega = (0, 1) \times (0, 1)$. Four meshes are used:

$$(3.9) \quad \text{2D Mesh } i := \text{1D Mesh } i \times \text{1D Mesh } i, \quad i = 1, 2, 3, 4.$$

On $\Pi_{COL} \times \Pi_{COL}$, the orthogonal Hermite bicubic collocation and five-point centered finite difference approximations to (3.8) are obtained in the same way as in the 1D case. They are also denoted (using the natural ordering) by $A_{COL}v = f$ and $A_F w = f$. A relation between the collocation and finite difference approximations can be obtained and expressed by

$$(3.10) \quad w = (B_C \otimes B_C)v,$$

where B_C is defined in (2.5) and \otimes denotes the Kronecker tensor product. Using A_F as a preconditioner, the resulting preconditioned collocation system is

$$(3.11) \quad T_2 w = f,$$

where

$$(3.12) \quad T_2 := A_F^{-1} (A_{COL}(B_C \otimes B_C)^{-1}).$$

The condition numbers of A_{COL} and T_2 are shown in Table 2 for both Dirichlet boundary conditions with $\alpha = 0$ and periodic boundary conditions with $\alpha = 1$. These results confirm the observations made in the 1D case. It is interesting to notice that for Mesh 4 the condition number again increases linearly with the measure β of its smoothness for both Dirichlet and periodic boundary conditions.

TABLE 2
Condition numbers of the collocation and preconditioned matrices for the 2D model problems.

MESH NO.	N	$\kappa_1(A_{COL})$ Dirichlet	$\kappa_1(T_2)$ ($\alpha = 0$)	$\kappa_1(A_{COL})$ Periodic	$\kappa_1(T_2)$ ($\alpha = 1$)	h_1^*	h_2^*
1 uniform	4	0.139E3	3.125	0.920E2	2.194	1.0	1.0
	8	0.583E3	3.128	0.399E3	2.196	1.0	1.0
	16	0.239E4	3.128	0.165E4	2.196	1.0	1.0
2	4	0.215E4	3.045	0.494E3	4.531	4.0	9.0
	8	0.351E5	3.082	0.697E4	9.059	4.0	49.0
	16	0.563E6	3.135	0.132E6	18.17	4.0	225.0
3	4	0.566E6	3.557	0.111E6	0.201E3	16.0	81.0
	8	0.145E9	3.622	0.258E8	0.206E4	16.0	2461.0
	16	0.371E11	3.761	0.804E10	0.189E5	16.0	50625.0
$\beta = 10$	8	0.554E5	3.743	0.224E5	3.238	8.0	10.0
	16	0.221E6	3.193	0.709E5	3.234	8.0	10.0
$\beta = 100$	8	0.552E7	25.08	0.237E7	25.85	80.0	100.0
	16	0.221E8	25.71	0.740E7	25.80	80.0	100.0
$\beta = 1000$	8	0.552E9	245.4	0.242E9	253.0	800.0	1000.0
	16	0.221E10	251.5	0.761E9	252.5	800.0	1000.0

3.3. Incomplete LU factorization preconditioning. For 1D problems, finite difference preconditioning is a quite inexpensive part of the iterative scheme. In higher dimensions, however, inversion of these finite difference matrices can become relatively expensive and/or complicated. This disadvantage is overcome by using a multigrid method or an incomplete LU factorization technique. Here, for simplicity we restrict our attention to the latter and the approach for a Dirichlet problem.

If a five-point finite difference approximation to a 2D second-order elliptic differential equations is given by

$$(3.13) \quad (AFu)_{i,j} = S_{i,j}u_{i,j-1} + W_{i,j}u_{i-1,j} + C_{i,j}u_{i,j} + N_{i,j}u_{i,j+1} + E_{i,j}u_{i+1,j},$$

then a five-diagonal incomplete LU factorization can be expressed by

$$(3.14) \quad \begin{aligned} (Lu)_{i,j} &= c_{i,j}u_{i,j} + s_{i,j}u_{i,j-1} + w_{i,j}u_{i-1,j}, \\ (Uu)_{i,j} &= u_{i,j} + n_{i,j}u_{i,j+1} + e_{i,j}u_{i+1,j}, \end{aligned}$$

where

$$(3.15) \quad \begin{aligned} s_{i,j} &= S_{i,j}, \\ w_{i,j} &= W_{i,j}, \\ c_{i,j} &= C_{i,j} - w_{i,j}e_{i-1,j} - s_{i,j}n_{i,j-1} - \gamma(s_{i,j}e_{i,j-1} + w_{i,j}n_{i-1,j}), \\ n_{i,j} &= N_{i,j}/c_{i,j}, \\ e_{i,j} &= E_{i,j}/c_{i,j} \end{aligned}$$

(see [CHQZ88] and [Won87]). The choice $\gamma = 0$ gives the standard incomplete LU factorization H_{LU5} (ILU) and $\gamma = 1$ gives the row sum equivalence version H_{RSS} (MILU [CE88]). Using these preconditionings for the A_F factor, we have the two new preconditioned matrices

$$T_3 := H_{RSS}^{-1} (A_{COL}(B_C \otimes B_C)^{-1}) \text{ and } T_4 := H_{LU5}^{-1} (A_{COL}(B_C \otimes B_C)^{-1}).$$

Note that the LU factorizations of B_C and $B_C \otimes B_C$ are inexpensive because of their sparse and banded structure.

TABLE 3

Extreme eigenvalues of the H_{RS5} and H_{LU5} preconditioned matrices for the 2D Laplacian operator with Dirichlet boundary conditions and $\alpha = 0$.

MESH NO.	N	T_3			T_4		
		$\min \lambda_i $	$\max \lambda_i $	$\max \left \frac{Im(\lambda_i)}{Re(\lambda_i)} \right $	$\min \lambda_i $	$\max \lambda_i $	$\max \left \frac{Im(\lambda_i)}{Re(\lambda_i)} \right $
1 uniform	4	1.197	3.728	0.0	0.389	3.040	0.0
	8	1.089	5.014	0.0	0.119	3.037	0.0
	16	1.044	9.814	0.0	0.032	3.037	0.0
2	4	1.133	3.365	0.121	0.590	3.047	0.106
	8	1.054	4.217	0.132	0.202	3.045	0.123
	16	1.025	5.894	0.163	0.054	3.073	0.141
3	4	1.178	3.717	0.426	0.949	3.686	0.623
	8	1.075	3.717	0.429	0.426	3.686	0.434
	16	1.032	5.031	0.429	0.126	3.686	0.434

Table 3 summarizes how the extreme eigenvalues of T_3 and T_4 depend on N and the smoothness of the mesh. There are some differences from using an exact factorization of the original finite difference matrix. Roughly speaking, the incomplete LU case is somewhat better for the high-frequency components and the row sum equivalence alternative is more accurate for the low-frequency components, as can be seen from Table 3. In fact, the minimal eigenvalue of T_3 is nearly one, which implies that H_{RS5} and $ACOL(B_C \otimes B_C)^{-1}$ have similar low-frequency components. On the other hand, the maximal eigenvalue of T_4 is bounded above by a constant independent of N , and H_{LU5} and $ACOL(B_C \otimes B_C)^{-1}$ have similar high-frequency components. Unlike the exact finite difference preconditioning, the condition numbers of both T_3 and T_4 can increase slowly with N . Since convergence rates for most conventional iterative methods are strongly related to the condition number, the row sum equivalence version on this basis appears to be more suitable as a preconditioning. Other important indicators for the convergence of iterative methods are ratios of real parts and imaginary parts of eigenvalues, which are listed in the table. It is obvious that the maximal ratio is strongly dependent upon the smoothness of meshes.

A more accurate factorization can be achieved by including one extra nonzero diagonal in each of L and U . The seven-diagonal LU factorization is given by (see [Won87] and [CHQZ88])

$$\begin{aligned}
 (Lu)_{i,j} &= c_{i,j}u_{i,j} + s_{i,j}u_{i,j-1} + w_{i,j}u_{i-1,j} + \tilde{e}_{i,j}u_{i+1,j-1}, \\
 (Uu)_{i,j} &= u_{i,j} + n_{i,j}u_{i,j+1} + e_{i,j}u_{i+1,j} + \tilde{w}_{i,j}u_{i-1,j+1},
 \end{aligned}
 \tag{3.16}$$

where the coefficients can be calculated from

$$\begin{aligned}
 s_{i,j} &= S_{i,j}, \\
 w_{i,j} &= W_{i,j} - s_{i,j}\tilde{w}_{i,j-1}, \\
 \tilde{e}_{i,j} &= -s_{i,j}e_{i,j-1}, \\
 c_{i,j} &= C_{i,j} - w_{i,j}e_{i-1,j} - s_{i,j}n_{i,j-1} - \tilde{e}_{i,j}\tilde{w}_{i+1,j-1} - \gamma(e_{i+1,j-1}\tilde{e}_{i,j} + w_{i,j}\tilde{w}_{i-1,j}), \\
 n_{i,j} &= N_{i,j}/c_{i,j}, \\
 e_{i,j} &= (E_{i,j} - n_{i+1,j-1}\tilde{e}_{i,j})/c_{i,j}, \\
 \tilde{w}_{i,j} &= -w_{i,j}n_{i-1,j}/c_{i,j}.
 \end{aligned}
 \tag{3.17}$$

Now $\gamma = 0$ gives the H_{LU7} version and $\gamma = 1$ the H_{RS7} version. Table 4 shows the extreme

TABLE 4

Extreme eigenvalues of the H_{RS7} and H_{LU7} preconditioned matrices for the 2D Laplacian operator with Dirichlet boundary conditions and $\alpha = 0$.

MESH NO.	N	T_5			T_6		
		$\min \lambda_i $	$\max \lambda_i $	$\max \left \frac{Im(\lambda_i)}{Re(\lambda_i)} \right $	$\min \lambda_i $	$\max \lambda_i $	$\max \left \frac{Im(\lambda_i)}{Re(\lambda_i)} \right $
1 uniform	4	1.168	3.102	0.0	0.714	3.037	0.0
	8	1.082	3.280	0.0	0.278	3.020	0.0
	16	1.043	5.458	0.0	0.081	3.018	0.0
2	4	1.103	3.102	0.135	0.901	3.047	0.113
	8	1.049	3.215	0.142	0.433	3.031	0.130
	16	1.010	3.570	0.144	0.134	3.016	0.144
3	4	1.170	3.672	0.574	1.140	3.671	0.543
	8	1.070	3.672	0.425	0.743	3.671	0.425
	16	1.026	3.672	0.425	0.287	3.671	0.425

eigenvalues of $T_5 := H_{RS7}^{-1} A_{COL}(B \otimes B)^{-1}$ and $T_6 := H_{LU7}^{-1} A_{COL}(B \otimes B)^{-1}$. Results are consistent with the observation concerning T_3 and T_4 , but the condition numbers are smaller.

4. Condition number estimates. The numerical results in the previous sections illustrate that the eigenvalues of the preconditioned collocation systems are bounded below by one and above by a constant which is independent of the number of mesh points, except in the case of Mesh 2 and Mesh 3 with periodic boundary conditions. Here, we shall present a theoretical analysis of the preconditioned system. The discussion is restricted to the simple model problem (2.1) with periodic boundary conditions and a uniform collocation mesh (Mesh 1).

For a uniform collocation mesh $\{x_i\}_{i=0}^{2N}$ a corresponding finite difference approximation for (2.1) with periodic boundary conditions is produced on the mesh $\{x_i^*\}_{i=0}^{4N+1}$ defined in (2.3). Arranging the variables and equations to correspond to the node ordering $\{x_0^*, x_2^*, \dots, x_{4N}^*, x_1^*, \dots, x_{4N+1}^*\}$ and using the periodic boundary conditions $u(x_0^*) = u(x_{4N}^*)$ and $u(x_{4N+1}^*) = u(x_1^*)$, the global finite difference matrix is given by

$$(4.1) \quad A_F = \begin{pmatrix} (b + \alpha)I & -C \\ -C^T & (b + \alpha)I \end{pmatrix},$$

where I is the $2N \times 2N$ identity matrix,

$$(4.2) \quad C = \begin{pmatrix} c & 0 & \dots & a \\ a & c & \dots & 0 \\ & & \ddots & \ddots \\ & & & a & c \end{pmatrix}$$

is a circulant matrix,

$$(4.3) \quad a = \frac{2}{h_2(h_1 + h_2)}, \quad b = \frac{2}{h_1 h_2}, \quad c = \frac{2}{h_1(h_1 + h_2)},$$

$$(4.4) \quad h_1 = (1 - 2\sigma)h, \quad h_2 = 2\sigma h,$$

and h is the collocation mesh stepsize. The corresponding Hermite cubic collocation system is given by

$$A_C u + \alpha B_C u = f$$

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or

$$(4.5) \quad (A_C B_C^{-1} + \alpha I)\bar{u} = f$$

where $\bar{u} = B_C u$. The objective of this section is to establish bounds on condition numbers of the matrix

$$(4.6) \quad T = A_F^{-1}(A_C B_C^{-1} + \alpha I).$$

Our approach is an algebraic one.

4.1. Eigensystem of collocation matrix. The eigenvalues and eigenvectors for the finite difference matrix are well known, and the corresponding eigensystem for the Hermite collocation matrix has been analyzed by Dyksen [Dyk87]. He considers the generalized eigenvalue problem $-\bar{v}'' = \lambda \bar{v}$ and the corresponding matrix eigenvalue problem

$$(4.7) \quad A_C B_C^{-1} v = \lambda v$$

with Dirichlet boundary conditions and gives an analytic expression of the eigenvalues. The problem with periodic boundary conditions also has been analyzed in [Sun94b], and these results can be summarized as follows:

LEMMA 4.1. *Let (A_C, B_C) denote a Hermite cubic collocation discretization for (2.1) with periodic boundary conditions. The eigenvalue problem $A_C B_C^{-1} v = \lambda u$ with $\alpha = 0$ has a complete set of eigenvectors for which the eigenvalues are nonnegative and are given by*

$$(4.8) \quad \lambda_{\pm l}^c = \frac{-v \pm \sqrt{v^2 - 4\mu\omega}}{-2\mu}, \quad l = 1, \dots, 2N - 1, l \neq N, l \neq -N$$

and

$$(4.9) \quad \lambda_{-0}^c = \frac{24}{(1 - 4\tau^2)h^2}, \quad \lambda_0^c = 0, \quad \lambda_{-N} = \frac{12}{h^2}, \quad \lambda_N = \frac{9}{h^2},$$

where

$$\begin{aligned} \mu &= h^4[(16\tau^4 - 16\tau^2 + 3)\rho - 8\tau^2 + 2], \\ v &= h^2[(-128\tau^2 + 48)\rho + 48], \\ \omega &= 192\rho, \\ \rho &= \tan^2\left(\frac{\theta_l}{2}\right), \quad \theta_l = l\pi/N, \end{aligned}$$

and $\tau = \frac{1}{2\sqrt{3}}$ ($\tau = \frac{1}{2} - \sigma$) for Gaussian collocation points. The corresponding eigenvectors are given by

$$(4.10) \quad v_l = d_l \begin{pmatrix} \left(-\frac{z_l}{2} \sin k\bar{\theta}_l - \tau h \cos k\bar{\theta}_l\right)_{k=0}^{2N-1} \\ \left(-\frac{z_l}{2} \sin k\bar{\theta}_l + \tau h \cos k\bar{\theta}_l\right)_{k=0}^{2N-1} \end{pmatrix}, \quad l = -0, 1 \leq |l| \leq N - 1,$$

$$(4.11) \quad v_l = d_l \begin{pmatrix} \left(\frac{z_l}{2} \cos k\bar{\theta}_l - \tau h \sin k\bar{\theta}_l\right)_{k=0}^{2N-1} \\ \left(\frac{z_l}{2} \cos k\bar{\theta}_l + \tau h \sin k\bar{\theta}_l\right)_{k=0}^{2N-1} \end{pmatrix}, \quad N + 1 \leq |l| \leq 2N - 1,$$

$$(4.12) \quad v_0 = \frac{1}{2} \begin{pmatrix} e \\ e \end{pmatrix}, \quad v_N = \frac{1}{2} \begin{pmatrix} -e^* \\ -e^* \end{pmatrix} \quad \text{and} \quad v_{-N} = \frac{1}{2} \begin{pmatrix} -e^* \\ e^* \end{pmatrix},$$

where

$$(4.13) \quad z_l = \frac{4}{h} \left(-\frac{1}{\lambda_l} - \frac{\tau^2 h^2}{6} + \frac{h^2}{8} \right) \tan \frac{\bar{\theta}_l}{2},$$

$$\bar{\theta}_l = |\theta_l|, \quad \theta_{-0} = \theta_0,$$

$e = (1, 1, \dots, 1)^T, e^* = (1, -1, 1, -1, \dots, 1, -1)^T \in \mathbb{R}^{2N}$ and d_l is a constant.

It is not difficult to obtain from Lemma 4.1 that

$$(4.14) \quad \pi^2 < \lambda_l^c \leq \frac{9}{h^2} \quad \text{and} \quad \frac{12}{h^2} < \lambda_{-l}^c < \lambda_{-0} = \frac{36}{h^2}, \quad l = 1, 2, \dots, 2N - 1.$$

The following Lemma is a result of the fact $A_C B_C^{-1}$ is symmetric when Gaussian points are chosen as the collocation points [Ben91].

LEMMA 4.2. *The eigenvectors $\{v_l\}_{l=\pm 0}^{\pm(2N-1)}$ defined in (4.10)–(4.12) form an orthogonal set under the inner product*

$$\langle u, v \rangle = \sum_i u_i v_i, \quad u, v \in \mathbb{R}^{2N}.$$

Close scrutiny reveals a natural bound for the ratios of eigenvalues of $A_C B_C^{-1}$ and the five-point finite difference matrix.

LEMMA 4.3. *The eigenvalues $\lambda_l^c = \lambda_l(A_C B_C^{-1}), l = \pm 0, \pm 1, \dots, \pm(2N - 1)$ defined in (4.8) and (4.9) satisfy*

$$(4.15) \quad 4 \leq \frac{h^2 \lambda_l^c}{\sin^2 \frac{\theta_l}{2}} \leq 9, \quad l = 1, 2, \dots, 2N - 1.$$

Proof. It follows from Lemma 4.1 that for $1 \leq l \leq 2N - 1$,

$$(4.16) \quad \frac{h^2 \lambda_l^c}{\sin^2 \frac{\theta_l}{2}} = \frac{(7\rho + 9) - \sqrt{\rho^2 + 90\rho + 81}}{(\frac{2}{3}\rho + \frac{1}{2}) \sin^2 \frac{\theta_l}{2}}.$$

Letting

$$G \left(\sin^2 \frac{\theta_l}{2} \right) := \frac{(7\rho + 9) - \sqrt{\rho^2 + 90\rho + 81}}{(\frac{2}{3}\rho + \frac{1}{2}) \sin^2 \frac{\theta_l}{2}},$$

where $\rho = \frac{\sin^2 \frac{\theta_l}{2}}{1 - \sin^2 \frac{\theta_l}{2}}$, for $0 < x < 1$ we have

$$G'(x) = -\frac{\sqrt{81 - 72x - 8x^2}(\frac{2}{3}x^2 - 6x - 9) + (-\frac{8}{3}x^3 - 36x^2 + 18x + 81)}{2(\frac{1}{6}x^2 + \frac{1}{2}x)^2 \sqrt{81 - 72x - 8x^2}} \geq 0.$$

Thus,

$$\max_{1 \leq l \leq 2N-1} \frac{h^2 \lambda_l^c}{\sin^2 \frac{\theta_l}{2}} \leq \lim_{x \rightarrow 1} G(x) = 9$$

and

$$\min_{1 \leq l \leq 2N-1} \frac{h^2 \lambda_l^c}{\sin^2 \frac{\theta_l}{2}} \geq \lim_{x \rightarrow 0} G(x) = 4. \quad \square$$

From the above proof, we can see that the bounds in (4.15) are asymptotically optimal. Let

$$(4.17) \quad w_l^s := (\sin k\theta_l)_{k=0}^{2N-1}, \quad \text{and} \quad w_l^c := (\cos k\theta_l)_{k=0}^{2N-1}, \quad l = 0, 1, 2, \dots, 2N - 1,$$

be $2N$ -dimensional vectors. If A is a circulant matrix defined by

$$A = \begin{pmatrix} a_0 & a_1 & & a_p \\ a_p & a_0 & a_1 & a_{p-1} \\ \dots & \dots & & \\ a_1 & a_2 & & a_p & a_0 \end{pmatrix}$$

then it is well known [Dav79] that its eigenvalues are given by

$$(4.18) \quad \lambda_j(A) = f(\epsilon_j), \quad j = 0, 1, \dots, p,$$

where

$$f(x) := a_0 + a_1 x + \dots + a_p x^p, \quad \epsilon_j := e^{\frac{i2\pi j}{p+1}}.$$

Furthermore, A can be decomposed into

$$(4.19) \quad A = F \Lambda(A) F^*,$$

where

$$F := (\epsilon_j^k)_{j,k=0}^p \quad \text{and} \quad \Lambda(A) := \text{diag}(1, f(\epsilon_1), \dots, f(\epsilon_p)).$$

LEMMA 4.4. a) *If the $2N \times 2N$ circulant matrix A is symmetric then*

$$(4.20) \quad \begin{aligned} Aw_j^c &= \lambda_j(A) w_j^c \\ Aw_j^s &= \lambda_j(A) w_j^s. \end{aligned}$$

b) *If A is skew symmetric then*

$$(4.21) \quad \begin{aligned} Aw_j^c &= -\gamma_j w_j^s \\ Aw_j^s &= \gamma_j w_j^c, \end{aligned}$$

where $\gamma_j = \text{Im}(f(\epsilon_j))$.

Proof. Lemma 4.4a) is a straightforward calculation, noting the fact that

$$F_j = w_j^c + i w_j^s,$$

where F_j is the j th column vector of F . By (4.19)

$$A(w_j^c + i w_j^s) = f(\epsilon_j)(w_j^c + i w_j^s).$$

Since each eigenvalue of a skew-symmetric circulant matrix is imaginary, $f(\epsilon_j) = i\gamma_j$, so (4.21) follows. \square

Let

$$(4.22) \quad \begin{aligned} u_l &= \begin{pmatrix} w_l^s \\ 0 \end{pmatrix}, & u_{-N-l} &= \begin{pmatrix} 0 \\ w_{N+l}^s \end{pmatrix}, & l &= 1, 2, \dots, N-1, \\ u_{-l} &= \begin{pmatrix} 0 \\ w_l^c \end{pmatrix}, & u_{N+l} &= \begin{pmatrix} -w_{N+l}^c \\ 0 \end{pmatrix}, & l &= 0, 1, \dots, N-1, \\ u_0 &= \begin{pmatrix} e \\ 0 \end{pmatrix}, & \text{and} & & u_{-N} &= \begin{pmatrix} 0 \\ e^* \end{pmatrix}. \end{aligned}$$

It follows directly that the eigenvectors of $A_C B_C^{-1}$ defined in (4.10)–(4.12) can be expressed by

$$(4.23) \quad (v_l, v_{-l}) = \tilde{I}(u_l, u_{-l})D_l,$$

where

$$(4.24) \quad \tilde{I} = \frac{1}{2} \begin{pmatrix} I & -I \\ I & I \end{pmatrix}, \quad D_l = \begin{pmatrix} -z_l d_l & -z_{-l} d_{-l} \\ 2\tau h d_l & 2\tau h d_{-l} \end{pmatrix}, \quad l = 1, \dots, 2N-1, l \neq N,$$

$$(4.25) \quad D_0 = D_N = I_2 \equiv \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

with

$$(4.26) \quad d_l = \sqrt{\frac{-z_{-l}}{4\tau^2 h^2 (z_l - z_{-l})}}, \quad d_{-l} = \sqrt{\frac{z_l}{4\tau^2 h^2 (z_l - z_{-l})}}, \quad l = 1, 2, \dots, (2N-1), l \neq N.$$

LEMMA 4.5. *Each $D_l, l = 0, 1, \dots, 2N-1$, is an orthogonal matrix.*

Proof. It is obvious that $D_0 D_0^T = D_N D_N^T = I_2$. For $l = 1, 2, \dots, (2N-1), l \neq N$, by (4.24) we obtain

$$D_l D_l^T = \begin{pmatrix} z_l^2 d_l^2 + z_{-l}^2 d_{-l}^2 & d_l d_{-l} (z_l z_{-l} + 4\tau^2 h^2) \\ d_l d_{-l} (z_l z_{-l} + 4\tau^2 h^2) & 4\tau^2 h^2 (d_l^2 + d_{-l}^2) \end{pmatrix}.$$

It follows from (4.8) and (4.13) that

$$z_l z_{-l} = \frac{16\rho}{h^2} \left(-\frac{1}{\lambda_l} + \frac{h^2}{9} \right) \left(-\frac{1}{\lambda_{-l}} + \frac{h^2}{9} \right) = -\frac{h^2}{3},$$

so

$$(4.27) \quad z_l z_{-l} + 4\tau^2 h^2 = 0.$$

Using (4.26) and (4.27), we obtain

$$z_l^2 d_l^2 + z_{-l}^2 d_{-l}^2 = 4\tau^2 h^2 (d_l^2 + d_{-l}^2) = 1.$$

Thus,

$$D_l D_l^T = I_2. \quad \square$$

4.2. Bounds for condition numbers of preconditioned systems. The key to proving uniform boundedness of condition numbers of the preconditioned matrices T in (4.6) is to show that $A_C B_C^{-1}$ and the preconditioner A_F share 2D invariant subspaces.

THEOREM 4.1. *Let (A_C, B_C) define a Hermite cubic collocation discretization of (2.1) and (2.3) and $\{v_l\}_{l=\pm 0}^{\pm(2N-1)}$, defined in (4.10)–(4.12), be the eigenvectors of $A_C B_C^{-1}$. Then each $S_l = \text{span}\{v_l, v_{-l}\}$, $l = 0, 1, \dots, 2N - 1$, is an invariant subspace of A_F .*

Proof. By (4.1) and (4.24), we have

$$(4.28) \quad \tilde{I}^T A_F \tilde{I} = \frac{1}{2} \begin{pmatrix} (b + \alpha)I - \frac{1}{2}(C + C^T) & \frac{1}{2}(C^T - C) \\ \frac{1}{2}(C - C^T) & (b + \alpha)I + \frac{1}{2}(C + C^T) \end{pmatrix},$$

where

$$C + C^T = \begin{pmatrix} 2c & a & & a \\ a & 2c & a & \\ \dots & & & \\ a & & a & 2c \end{pmatrix}, \quad C - C^T = \begin{pmatrix} 0 & -a & & a \\ a & 0 & -a & \\ \dots & & & \\ -a & & a & 0 \end{pmatrix}.$$

Thus,

$$A_F(v_l, v_{-l}) = \tilde{I} \begin{pmatrix} (b + \alpha)I - \frac{1}{2}(C + C^T) & \frac{1}{2}(C^T - C) \\ \frac{1}{2}(C - C^T) & (b + \alpha)I + \frac{1}{2}(C + C^T) \end{pmatrix} (u_l, u_{-l}) D_l.$$

It follows from Lemma 4.4 that

$$(4.29) \quad \begin{cases} \frac{1}{2}(C + C^T)w_j^s = (c + a \cos \theta_l)w_j^s \\ \frac{1}{2}(C + C^T)w_j^c = (c + a \cos \theta_l)w_j^c, \end{cases} \quad \begin{cases} \frac{1}{2}(C - C^T)w_j^s = -a \sin \theta_l w_j^c \\ \frac{1}{2}(C - C^T)w_j^c = a \sin \theta_l w_j^s, \end{cases}$$

and therefore,

$$(4.30) \quad A_F(v_l, v_{-l}) = \tilde{I}(u_l, u_{-l}) \begin{pmatrix} b + \alpha - c - a \cos \theta_l & -a \sin \theta_l \\ -a \sin \theta_l & b + \alpha + c + a \cos \theta_l \end{pmatrix} D_l$$

$$l = 0, 1, \dots, 2N - 1.$$

Letting

$$(4.31) \quad \bar{D}_l := \begin{pmatrix} (b + \alpha) - (c + a \cos \theta_l) & -a \sin \theta_l \\ -a \sin \theta_l & (b + \alpha) + (c + a \cos \theta_l) \end{pmatrix}$$

and using Lemma 4.5 and equation (4.23), (4.30) can be rewritten as

$$(4.32) \quad A_F(v_l, v_{-l}) = (v_l, v_{-l}) D_l^T \bar{D}_l D_l.$$

Thus, S_l , $l = 0, 1, \dots, 2N - 1$, are invariant subspaces of A_F . \square

It remains to estimate the spectrum in each subspace S_l . The basic result is given in the theorem below; its detailed proof is in [SHR93].

THEOREM 4.2. *Given (A_C, B_C) and A_F as in Theorem 4.1 with $\alpha > 0$, there exist positive constants C_1 and C_2 independent of h such that T defined in (4.6) satisfies*

$$(4.33) \quad 0 < C_1 \leq \lambda_l(T^T T) \leq C_2, \quad l = 0, 1, \dots, 2N - 1.$$

Let $\kappa_2(T)$ denote the condition number of T for the L_2 norm, i.e.,

$$\kappa_2(T) = \frac{\max_i \lambda_i(T^T T)}{\min_i \lambda_i(T^T T)}.$$

The two corollaries below follow immediately.

COROLLARY 4.1. *There exists a constant C_1^* independent of h such that*

$$(4.34) \quad \kappa_2(T) \leq C_1^*.$$

COROLLARY 4.2. *There exist constants C_3^* and C_4^* independent of h such that*

$$(4.35) \quad 0 < C_3^* \leq \lambda_l(T) \leq C_4^*, \quad l = 0, 1, \dots, 2N - 1,$$

and

$$(4.36) \quad \kappa_1(T) \leq \frac{C_4^*}{C_3^*}, \quad l = 0, 1, \dots, 2N - 1.$$

Remarks. (i) Since each eigenvalue of $T = A_F^{-1}(A_C B_C^{-1} + \alpha I)$ is a continuous function of α in $(0, \infty)$ with

$$\lim_{\alpha \rightarrow \infty} \lambda_l(T) = 1 \quad \text{and} \quad \lim_{\alpha \rightarrow 0} \lambda_l(T) < \text{constant}, \quad l = 0, 1, \dots, 2N - 1,$$

C_1 and C_2 can be assumed to be independent of α .

(ii) Theorem 4.2, while showing that the eigenvalues of the preconditioned matrix T are bounded by constants, does not give an explicit bound. Numerical observation has shown that each $\lambda_i(T)$ decreases as α increases, with $\min \lambda_i(T) = \lim_{\alpha \rightarrow \infty} \lambda_i(T) = 1$. Furthermore, for a given $\alpha > 0$, the eigenvectors corresponding to both maximum and minimum eigenvalues of T occur in the subspace S_0 . A straightforward calculation gives [SHR93]

$$\min \lambda_i(T) = 1 \quad \text{and} \quad \max \lambda_i(T) = 18\tau - 3 + o(h) \approx 2.19615 + o(h),$$

which implies $\kappa_1(T)$ is approximately bounded by 2.196. This agrees with the numerical results in §2 (see Table 2). The same relationships also hold in the case $\kappa_2(T)$.

When $\alpha = 0$, A_F is singular, so T is undefined. We can then consider a generalized eigenvalue problem $A_C B_C^{-1} v = \lambda A_F v$. The eigenvalues corresponding to the subspace S_l ($l > 0$) lie in $[1, 18\tau - 3]$.

(iii) One can use central differences with a uniform mesh on $[0, 1]$ instead of the nonuniform mesh $\{x_i^*\}$ in (2.3) and obtain a matrix having the same form as (4.5) with

$$b = \frac{2}{(2N + 1)^2} \quad \text{and} \quad a = c = \frac{1}{(2N + 1)^2}.$$

The above arguments hold with minor modification, and the boundedness of the eigenvalues of T and $T^T T$ is still valid.

(iv) It is easy to find examples of some discretizations of PDEs for which the spectral condition numbers are bounded but the condition numbers in the L_2 -norm are not. Nevertheless, some iterative methods can still be very effective and converge quickly when applied to such preconditioned systems if the eigenvalues are properly distributed.

It should be noted that the collocation system preconditioned by a simple finite difference scheme is a very special case. Since A_F and $A_C B_C^{-1}$ have $2N$ common invariant subspaces for a uniform mesh and periodic boundary conditions, both spectral condition numbers and

condition numbers in the L_2 -norm are bounded by constants independent of h . Our numerical results indicate that in the case of a nonuniform mesh and/or Dirichlet boundary conditions, the spectral condition numbers are bounded, although we have as yet been unable to provide a theoretical proof. Numerical results indicate that the corresponding condition numbers in the L_2 -norm are unbounded [SHR93], however.

(v) Although the discussion is based on a 1D model problem, these theoretical results are straightforward to generalize for the corresponding 2D model problem.

5. Numerical solution of PDEs. The collocation matrices A_C and B_C are generally nonsymmetric, as are the resulting preconditioned matrices. A number of iterative methods have been developed in recent years for the solution of nonsymmetric preconditioned systems. They would be expected to be efficient in our context because of the relatively small condition numbers and the close concentration of eigenvalues for the preconditioned matrices. To illustrate that this is indeed the case, we consider two examples of 2D second-order elliptic PDEs defined on $\Omega = [0, 2] \times [0, 2]$ with Dirichlet boundary conditions. The orthogonal Hermite bicubic collocation method is used for the discretization of the PDEs. In order to avoid the effect of discretization errors in our tests, the right-hand sides and boundary conditions in the two examples are chosen such that

$$(5.1) \quad u(x, y) = xy(2 - x)(2 - y)$$

is both the PDE solution and the collocation solution. The five-diagonal row sum equivalence incomplete LU factorization (H_{RSS}) is a preconditioner and a generalized conjugate residual (GCR) method [EES83] is employed to solve the preconditioned collocation equations. The error is measured by

$$(5.2) \quad e^{(k)} = \max_{ij} | u_{ij}^{(k)} - u_{ij} |,$$

where u_{ij} denotes the exact solution and its first and mixed derivative values at mesh points, and $u_{ij}^{(k)}$ denotes the corresponding numerical solution at the k th iteration. For both examples, the initial guess is taken as zero. The iterative procedure is stopped when $e^{(k)}$ is smaller than a prescribed tolerance δ .

Example 5.1 Poisson Equation.

$$(5.3) \quad \Delta u = f(x, y).$$

In Table 5, we list the number of iterations required for preconditioned GCR (PGCR) with four tolerances and using the 2D Meshes 1–3. Since this example is separable, other iterative methods such as block Gauss–Seidel (BGS), block SOR (BSOR) [Sun94a], and ADC [Dyk87] are also applicable when a uniform mesh is used. The results with BGS and BSOR are also given in Table 5. ADC is not considered since it is actually a direct method for this simple problem [Dyk87]. Although the number of iterations for PGCR to achieve a prescribed tolerance increases with N , the growth rate is acceptable and is smaller than that of BGS or BSOR. For given N , PGCR requires nearly the same number of iterations for 2D Mesh 2 and 3 as for a uniform mesh.

Example 5.2 A Nonseparable Elliptic PDE (Prob. 18 in [RB85]).

$$(5.4) \quad \frac{\partial^2 u}{\partial x^2} + (1 + xy) \frac{\partial^2 u}{\partial y^2} + \cos(x) \frac{\partial u}{\partial x} - e^{-x} \frac{\partial u}{\partial y} + 3u = f(x, y).$$

None of other iterative methods are applicable for this nonseparable problem. Numbers of iterations for PGCR using various meshes and tolerances are summarized in Table 6. The same basic conclusions about PGCR can be made as in the previous example.

TABLE 5

Example 5.1 Iteration numbers of (H_{RSS}) preconditioned GCR and (nonpreconditioned) BGS, BSOR methods.

METHOD	MESH NO.	N	$\delta = 10^{-3}$	$\delta = 10^{-4}$	$\delta = 10^{-5}$	$\delta = 10^{-6}$
PGCR	1	8	10	12	15	18
		16	16	19	22	24
	2	8	10	13	15	17
		16	17	19	22	24
	3	8	11	13	15	17
		16	22	26	29	31
BGS	1	8	22	29	37	44
		16	65	94	125	154
BSOR	1	8	9	13	15	16
		16	17	22	28	32

TABLE 6

Example 5.2 Iteration numbers of (H_{RSS}) preconditioned GCR method.

MESH NO.	N	$\delta = 10^{-3}$	$\delta = 10^{-4}$	$\delta = 10^{-5}$	$\delta = 10^{-6}$
1	4	9	11	12	14
	8	12	14	16	17
	16	16	19	21	24
2	4	8	10	12	13
	8	10	12	14	16
	16	16	19	23	25
3	4	7	10	11	13
	8	8	11	13	16
	16	23	27	30	32

6. Conclusions and comments. A technique has been presented to construct a finite difference preconditioning for orthogonal collocation matrices. Only cubic Hermite basis functions have been considered, although the preconditioning principles are straightforward to generalize for higher order collocation methods. It is shown numerically that the norms of eigenvalues of the preconditioned collocation matrices can be bounded below by one and above by a constant independent of the number of mesh nodes using the exact low-order finite difference preconditioner. For both smooth and moderately nonsmooth meshes, the preconditioned matrix has a small spectral condition number, with its eigenvalues located in a narrow interval on the positive real axis. As a consequence, conventional iterative methods such as Richardson and conjugate gradient iterations give fast convergence for the preconditioned collocation equations.

We have also investigated much cheaper alternatives involving incomplete LU factorizations based on a low-order finite difference approximation. The row sum equivalence version is particularly efficient although condition numbers may grow slowly with the number of mesh points. Moreover, the generalized conjugate residual method with this preconditioning has been seen to converge fast for both separable and nonseparable problems with uniform or nonuniform meshes.

While only limited experiments and comparisons have been performed, the results are quite encouraging. It is our intention to perform a theoretical analysis of these preconditioning schemes (possibly developing a general theoretical framework for the collocation methods analogous to what is done in [MP90] for the finite element method) and do a more careful comparison with the currently popular direct schemes for solving the collocation linear systems.

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