



Fast Numerical Solution of the Biharmonic Dirichlet Problem on Rectangles

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we concentrate on the discrete case and provide a description that is quite close to the computer algorithm, we note that a similar analysis is also possible for the continuous problem. In that case the analysis is related to the solution of the separable problem where Δu , instead of u_n , is specified on two opposite parts of the boundary. We will see in § 4 that the discrete analysis provides precise estimates of the rate of convergence. The matrix A can be represented with the aid of two simple matrices. Thus, let the negative of a one-dimensional discrete Laplace operator be the symmetric, positive definite, tridiagonal $N \times N$ matrix

$$(5) \quad R = \text{tridiag}[-1, 2, -1].$$

Let U be the $N \times 2$ matrix defined by

$$(6) \quad U = [e_1, e_N],$$

where e_i is the i th column of an $N \times N$ identity matrix I . The matrix A can be written

$$(7) \quad A = [(I \otimes R) + (R \otimes I)]^2 + 2(UU^T \otimes I) + 2(I \otimes UU^T).$$

The two last terms in (7) arise from the quadratic boundary extrapolation. The matrix

$$(8) \quad L = (I \otimes R) + (R \otimes I)$$

is the standard 5-point difference approximation of the negative Laplace operator in two dimensions. Let

$$(9) \quad B = L^2 + 2(UU^T \otimes I).$$

The matrix B represents a discrete approximation of the biharmonic problem with Δu specified rather than the normal derivative, on two opposite sides of the rectangle. For this problem separation of the variables is possible. We will show that this problem can be used in a special way, to precondition the original problem.

In the following let $P_{NM} \in R^{nm \times nm}$ be the permutation matrix such that if $D \in R^{m \times n}$, $E \in R^{k \times l}$ then

$$(10) \quad P_{KM}(D \otimes E)P_{LN}^T = (E \otimes D).$$

Notice that if the vector x with components x_{ij} , $i = 1, 2, \dots, M$, $j = 1, 2, \dots, N$ is defined on a grid with M rows and N columns, then $P_{MN}x$ is the permuted vector ordered along rows instead of columns. We also need the $N \times N$ orthogonal matrix

$$(11) \quad Q = \{q_{ij}\} = \sqrt{\frac{2}{N+1}} \left\{ \sin \frac{ij\pi}{N+1} \right\}.$$

It is easy to show that the vectors q_i , $i = 1, 2, \dots, N$ are the normalized eigenvectors of R and that

$$(12) \quad \begin{aligned} QRQ &= \Lambda, \\ Q &= Q^T = Q^{-1}, \\ \Lambda &= \text{diag}(\lambda_j), \\ \lambda_j &= 2 \left(1 - \cos \frac{j\pi}{N+1} \right), \quad j = 1, 2, \dots, N. \end{aligned}$$

Q represents a real sine transform and $y = Qx$ can be computed in $O(N \log N)$ operations using the fast Fourier transform.

Using the Sherman–Morrison formula [8],

$$(13) \quad B^{-1} = L^{-2}(I - 2(U \otimes I)\tilde{C}^{-1}(U^T \otimes I)L^{-2}),$$

where \tilde{C} is the $2N \times 2N$ matrix

$$(14) \quad \begin{aligned} \tilde{C} &= I + 2(U^T \otimes I)L^{-2}(U \otimes I) \\ &= I + 2((QU)^T \otimes I)\tilde{S}^{-1}(QU \otimes I) \end{aligned}$$

with

$$(15) \quad \tilde{S} = [(I \otimes R^2) + 2(\Lambda \otimes R) + (\Lambda^2 \otimes I)].$$

S is a block diagonal matrix, each block \tilde{S}_k , $k = 1, \dots, N$, being pentadiagonal.

THEOREM 1. *The solution of the linear system $\tilde{C}x = y$ can be reduced to the solution of the two linear systems*

$$\begin{aligned} \tilde{T}_1(x_1 + x_2) &= y_1 + y_2, \\ \tilde{T}_2(x_1 - x_2) &= y_1 - y_2, \end{aligned}$$

where $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$, $y = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$ and

$$\begin{aligned} \tilde{T}_1 &= I + \frac{8}{N+1} \sum_{k=1,3,\dots}^{N-1} \sin^2 \frac{k\pi}{N+1} \tilde{S}_k^{-1}, \\ \tilde{T}_2 &= I + \frac{8}{N+1} \sum_{k=2,4,\dots}^N \sin^2 \frac{k\pi}{N+1} \tilde{S}_k^{-1}. \end{aligned}$$

Proof. Performing the matrix multiplications in (14) gives

$$\tilde{C} = I + 2 \begin{pmatrix} \sum_{k=1}^N q_{k1}^2 \tilde{S}_k^{-1}, & \sum_{k=1}^N q_{k1} q_{kN} \tilde{S}_k^{-1} \\ \sum_{k=1}^N q_{k1} q_{kN} \tilde{S}_k^{-1}, & \sum_{k=1}^N q_{kN}^2 \tilde{S}_k^{-1} \end{pmatrix}.$$

Partition the equations according to this block structure and use $q_{kN} = (-1)^{k+1} q_{k1}$. The final result is then obtained by adding and subtracting the block equations. \square

Equation (13) shows that the solution of a linear system $Bx = y$ can be obtained if one can solve linear systems with coefficient matrices L and \tilde{C} . At most $O(N^2)$ operations are required to solve the linear system $\tilde{C}x = y$. This follows from Theorem 1 and the fact that QT, Q is diagonal. An important observation is that several $O(N^2)$ methods for solving the discrete Poisson equation (matrix L) are known; see, for example, Bank and Rose [1] and Schröder, Trottenberg and Witsch [25]. Such a method must be used in order to obtain an $O(N^2)$ method for the present problem. Direct specialized algorithms for the solution of linear systems with coefficient matrix B can also be devised. Alternatively, one can proceed using a method based on Fourier transforms, since

$$(16) \quad \begin{aligned} B &= (I \otimes Q)P_{NN}^T[(\Lambda^2 \otimes I) + 2(\Lambda \otimes R) + (I \otimes R^2) + 2(I \otimes UU^T)]P_{NN}(I \otimes Q) \\ &= (I \otimes Q)P_{NN}^T S P_{NN}(I \otimes Q), \end{aligned}$$

where

$$(17) \quad S = \tilde{S} + 2(I \otimes UU^T).$$

S is block diagonal and the k th block of S is

$$(18) \quad S_k = \tilde{S}_k + 2UU^T.$$

Using this decomposition, the solution takes $O(N^2 \log N)$ operations. Next, consider the formal inversion of the discrete biharmonic operator A

$$(19) \quad A^{-1} = B^{-1}(I - 2(I \otimes U)C^{-1}(I \otimes U^T)B^{-1})$$

where

$$(20) \quad C = I + 2(I \otimes U^T)B^{-1}(I \otimes U).$$

If an $O(N^2)$ algorithm for the solution of a linear system $Cx = y$ can be found, then the discrete biharmonic equation can be solved in $O(N^2)$ operations.

THEOREM 2. *The solution of the linear system $Cx = y$ can be reduced to the solution of the two linear systems,*

$$T_1(x_1 + x_2) = y_1 + y_2,$$

$$T_2(x_1 - x_2) = y_1 - y_2,$$

where

$$P_{2N}x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}, \quad P_{2N}y = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

and

$$T_1 = I + \frac{8}{N+1} \sum_{k=1,3}^{N-1} \sin^2 \frac{k\pi}{N+1} S_k^{-1},$$

$$T_2 = I + \frac{8}{N+1} \sum_{k=2,4}^N \sin^2 \frac{k\pi}{N+1} S_k^{-1}.$$

Proof. The proof is similar to that of Theorem 1 with \tilde{S} replaced by S and a permutation P_{2N} of the variables. \square

The next theorem provides the basis for the analysis of an iterative method for the solution of linear systems with coefficient matrix T_r , $r = 1, 2$.

THEOREM 3. *The following matrix has the block structure*

$$P_{2N}(Q\tilde{T}_r Q)^{-1/2} Q T_r Q (Q\tilde{T}_r Q)^{-1/2} P_{2N}^T = I - \begin{bmatrix} (F^{r1})^T F^{r1} & 0 \\ 0 & (F^{r2})^T F^{r2} \end{bmatrix}$$

for $r = 1, 2$. Moreover, if we let $\psi_{ij} = (\lambda_i + \lambda_j)^2$, $i_r = 2(i-1) + r$, $j_r = 2(j-1) + r$ and

$$\alpha_k^r = 1 + \frac{8}{N+1} \sum_{j=r, r+2, \dots}^N \sin^2 \frac{j\pi}{N+1} \Psi_{kj}^{-1},$$

then F^{rs} is the $N/2 \times N/2$ matrix with components

$$f_{ij}^{rs} = \frac{8}{N+1} \sin \frac{i_r \pi}{N+1} \sin \frac{j_s \pi}{N+1} \psi_{i_r j_s}^{-1} / \sqrt{\alpha_{i_r}^s \alpha_{j_s}^r}.$$

Proof. Write

$$\begin{aligned} Q T_r Q &= I + \frac{8}{N+1} \sum_{k=r, r+2, \dots}^N \sin^2 \frac{k\pi}{N+1} Q S_k^{-1} Q \\ &= I + \frac{8}{N+1} \sum_{k=r, r+2, \dots}^N \sin^2 \frac{k\pi}{N+1} \Psi_k^{-1} (I - QU(I_2 + 2(QU)^T \Psi_k^{-1} QU)^{-1} (QU)^T \Psi_k^{-1}), \end{aligned}$$

where $\Psi_k = \text{diag}(\Psi_{kj}), j = 1, 2, \dots, N$. It is clear that exactly the same kind of calculation that lead to Theorems 1 and 2 can be repeated, this time working with scalars. The calculation is straightforward but tedious. For more details we refer to [3]. \square

Observe that the two matrix problems associated with T_r ($r = 1, 2$) have been split into four smaller problems. This reduces the required computer storage since we can process one problem at a time. The reduction into four subproblems is a consequence of the symmetry of the biharmonic operator on the rectangle R . Each subproblem corresponds to a subspace of the space of biharmonic eigenfunctions. Consider the square $R = \{(x, y): |x| < 1, |y| < 1\}$. The discrete biharmonic eigenfunctions with symmetry around the coordinate axis and symmetry or antisymmetry around the diagonals are generated by the matrix F^{11} , while the eigenfunctions with antisymmetry around the coordinate axis and symmetry or antisymmetry with respect to the diagonals are generated by F^{22} . The matrices F^{12} and F^{21} generate eigenfunctions which are antisymmetric under a rotation of π . This is a degenerate case and for each eigenvalue in this group there are two eigenfunctions of the same shape, one rotated $\pi/2$ relative to the other. A second important observation is that the elements f_{ij}^{rs} can be computed easily after some preprocessing of the quantities that appear in the above formula. This requires only $O(N^2)$ operations and $O(N)$ storage and provides an alternative to the implicit definition of T_r given in Theorem 2.

3. A preconditioned conjugate gradient method. A very attractive iterative scheme for the solution of a symmetric positive definite linear system $Ax = b$ is the conjugate gradient method. From an arbitrary initial vector x_0 the method generates a sequence of approximations $\{x_n\}$ to the solution x defined by

$$\begin{aligned} x_{n+1} &= x_n + \alpha_n p_n, & \alpha_n &= (r_n, r_n) / (A p_n, p_n), \\ p_{n+1} &= r_{n+1} + \beta_n p_n, & \beta_n &= (r_{n+1}, r_{n+1}) / (r_n, r_n), \end{aligned}$$

where $r_n = b - Ax_n$ and $p_0 = r_0$. The method is due to Hestenes and Stiefel [17]. The iteration does not require knowledge of the matrix elements, since only matrix vector products are needed. It therefore follows from Theorem 2 that this iterative method can be used to solve the linear system $Cx = y$. It can be shown [3] that this method requires $O(N^{1/3})$ iterations resulting in an $O(N^{7/3})$ method for the biharmonic problem.

Suppose, instead of applying the conjugate gradient method directly to a matrix T , that we split T by writing

$$T = \tilde{T} - (\tilde{T} - T).$$

Assume that it is easy to solve linear systems with the matrix \tilde{T} . In this case the conjugate gradient method can be used with a preconditioning matrix \tilde{T} corresponding to the above splitting of T . An analysis of this technique is given in Concus, Golub and O'Leary [7]. The process can be viewed equivalently as applying an ordinary conjugate gradient iteration to the transformed system $\tilde{T}^{-1/2} T \tilde{T}^{-1/2}$ with a change of variables. If \tilde{T}^{-1} is an approximate inverse of T , then the convergence rate will be much improved. Two effects can contribute to this. First, the ratio between the largest and the smallest eigenvalue μ_{\max}/μ_{\min} is often substantially reduced when we consider $\tilde{T}^{-1/2} T \tilde{T}^{-1/2}$ instead of T . Second, and often more important, the spectrum of $\tilde{T}^{-1/2} T \tilde{T}^{-1/2}$ will usually have a more favorable distribution. Typically, there will be a cluster of eigenvalues close to one, and only a few outlying eigenvalues. We propose to solve the linear systems $T_r x = y$, using \tilde{T}_r as a preconditioning matrix. We will show in the next section that both of the above mentioned effects are prominent in this case.

4. Convergence of the iterative method. We will in this section prove properties about the spectrum of $\tilde{T}_r^{-1}T_r$. This enables us to determine the rate of convergence of the iterative method proposed in the previous section. It follows from Theorem 3 that the spectrum of $\tilde{T}_r^{-1}T_r$ can be studied by considering the singular values of the four matrices F^{rs} , $r = 1, 2, s = 1, 2$. The following lemma enables us to express the quantity α_k^r , defined in Theorem 3, in closed form.

LEMMA 1. For $0 < a < 1$,

$$4a^2 \sum_{j=1,2,\dots}^N \frac{\sin^2 \frac{j\pi}{N+1}}{\left(1+a^2-2a \cos \frac{j\pi}{N+1}\right)^2} = 2(N+1) \left\{ \frac{a^2}{1-a^2} - \frac{(a^{N+1})^2}{1-(a^{N+1})^2} \left(\frac{2(N+1)}{1-(a^{N+1})^2} - \frac{1+a^2}{1-a^2} \right) \right\}$$

and

$$4a^2 \sum_{j=2,4,\dots}^N \frac{\sin^2 \frac{j\pi}{N+1}}{\left(1+a^2-2a \cos \frac{j\pi}{N+1}\right)^2} = (N+1) \left\{ \frac{a^2}{1-a^2} - \frac{a^{N+1}}{1-a^{N+1}} \left(\frac{N+1}{1-a^{N+1}} - \frac{1+a^2}{1-a^2} \right) \right\}.$$

Proof. Let

$$f(x) = \frac{4a^2 \sin^2 x}{(1+a^2-2a \cos x)^2}.$$

Poisson's summation formula gives the relation

$$(21) \quad \frac{1}{2}f(0) + \sum_{k=1}^N f\left(\frac{k\pi}{N+1}\right) + \frac{1}{2}f(\pi) = \frac{N+1}{\pi} \left[F_0 + 2 \sum_{k=1}^{\infty} F_{2k(N+1)} \right],$$

where

$$F_k = \int_0^\pi f(x) \cos kx \, dx.$$

Integration by parts reduces this to well known integrals which can be found in [13]. When substituting the result back into (21), we are left with geometric series. The first result is obtained by performing the summations. The second result, where the sum extends over even integers only, follows in the same way by a change of variable. \square

The sum over odd integers can now be found as the difference between the two expressions in Lemma 1. Together these results furnish closed form expressions for the individual matrix elements f_{ij}^{rs} defined in Theorem 3. The elements are positive and increase with N . The following important lemma gives the precise form of the limit matrix as the dimension N becomes large.

LEMMA 2. The matrix F^{rs} defined in Theorem 3 has elements:

$$f_{ij}^{rs} = \frac{8}{\pi} \frac{(i_j^s)^{3/2}}{(i_r^2 + j_s^2)^2} \frac{a_i^{rs} a_j^{sr}}{b_i^{rs} b_j^{sr}} + O\left(\frac{1}{(N+1)^2}\right), \quad r = 1, 2, \quad s = 1, 2,$$

where

$$i_r = 2(i - 1) + r, \quad j_r = 2(j - 1) + r$$

and a_j^{rs} and b_j^{rs} are exponentially close to one in j and given by

$$a_j^{rs} = 1 + (-1)^{s-1} e^{-i_r \pi},$$

$$b_j^{rs} = (1 + 2(-1)^{s-1} j_r \pi e^{-i_r \pi} - e^{-2i_r \pi})^{1/2}.$$

Proof. Derive Taylor expansions for each element f_{ij}^{rs} in the variable $1/(N + 1)$ around zero. This is tedious to do by hand and the symbolic manipulation program MACSYMA [20] was used to derive the above expressions. \square

As an illustration of Lemma 2, the 3×3 leading principal minors of the (infinite) limit matrix F_∞^{rs} are compared with the corresponding minors of F_{63}^{rs} for $N = 63$ in Fig. 1. Notice that the approximation is quite good already for this value of N .

$F_{63}^{11} = \begin{bmatrix} .545 & .122 & .038 \\ .122 & .209 & .125 \\ .038 & .125 & .123 \end{bmatrix}$	$F_\infty^{11} = \begin{bmatrix} .546 & .122 & .039 \\ .122 & .212 & .128 \\ .039 & .128 & .127 \end{bmatrix}$
$F_{63}^{12} = \begin{bmatrix} .319 & .078 & .030 \\ .218 & .167 & .093 \\ .093 & .132 & .108 \end{bmatrix}$	$F_\infty^{12} = \begin{bmatrix} .320 & .079 & .031 \\ .219 & .169 & .096 \\ .095 & .135 & .112 \end{bmatrix}$
$F_{63}^{22} = \begin{bmatrix} .323 & .144 & .065 \\ .144 & .156 & .107 \\ .065 & .107 & .101 \end{bmatrix}$	$F_\infty^{22} = \begin{bmatrix} .325 & .146 & .067 \\ .146 & .159 & .111 \\ .067 & .111 & .106 \end{bmatrix}$

FIG. 1. Leading principal minors of F_N^{rs} for $N = 63$ and $N = \infty$.

The next lemma can be used to obtain bounds on the row and column sums of F_∞^{rs} .

LEMMA 3. *Let*

$$S_i^r = \sum_{j=1}^{\infty} \frac{(i_j^r)^{3/2}}{(i_r^2 + j_r^2)^2}, \quad r = 1 \text{ or } 2,$$

for some given $i \in (1, 2, 3, \dots)$. Then

$$\frac{\pi\sqrt{2}}{16} - \frac{1}{i} \frac{25}{32} \left(\frac{3}{5}\right)^{3/4} \leq S_i^r \leq \frac{\pi\sqrt{2}}{16} + \frac{1}{i} \frac{25}{32} \left(\frac{3}{5}\right)^{3/4}$$

for all positive i and $r = 1$ or 2 .

Proof. Let

$$S_i = \sum_{j=1}^{\infty} \frac{i^{3/2} j^{3/2}}{(i^2 + j^2)^2} = \frac{1}{i} \sum_{j=1}^{\infty} \frac{(j/i)^{3/2}}{(1 + (j/i)^2)^2}.$$

Consider

$$f(x) = \frac{x^{3/2}}{(1 + x^2)^2}$$

with

$$f_{\max} = f\left(\sqrt{\frac{3}{5}}\right) = \frac{25}{16} \left(\frac{3}{5}\right)^{3/4}, \quad 0 \leq x \leq \infty,$$

and

$$\int_0^\infty f(x) dx = \frac{\pi\sqrt{2}}{8}.$$

Clearly

$$\lim_{i \rightarrow \infty} S_i = \int_0^\infty f(x) dx = \frac{\pi\sqrt{2}}{8}.$$

By considering the discrete sum for finite i and the fact that f is monotone on each side of its maximum, it follows that

$$\frac{\pi\sqrt{2}}{8} - \frac{1}{i}f_{\max} \leq S_i \leq \frac{\pi\sqrt{2}}{8} + \frac{1}{i}f_{\max}.$$

Doing the same analysis for the even sum S_{even} ,

$$S_{\text{even}} = \frac{1}{i} \sum_{j=2,4,\dots}^\infty \frac{(j/i)^{3/2}}{(1+(j/i)^2)^2},$$

results in

$$\frac{\pi\sqrt{2}}{16} - \frac{1}{i}g_{\max} \leq S_{\text{even}} \leq \frac{\pi\sqrt{2}}{16} + \frac{1}{i}g_{\max}$$

where the appropriate function is

$$g(x) = \frac{2\sqrt{2}x^{3/2}}{(1+4x^2)^2}, \quad \int_0^\infty g(x) dx = \frac{\pi\sqrt{2}}{16}$$

and

$$g_{\max} = g\left(\sqrt{\frac{3}{20}}\right) = f_{\max}.$$

Combining these two results proves the lemma. \square

We can now give the following bound on the singular values of the matrices F^{rs} .

THEOREM 4. *Let $\{\sigma_i\}_{i=1}^N$ be the singular values of one of the matrices F^{rs} defined in Theorem 3. Then*

$$0 \leq \sigma_i < 0.8,$$

independent of N .

Proof. An upper bound for the largest singular value σ_1 of the matrices F^{rs} will be derived. The following elementary inequality will be used:

$$\begin{aligned} \sigma_1 &\leq \|(F^{rs})^T F^{rs}\|_\infty^{1/2} \leq (\|F^{rs}\|_1 \|F^{rs}\|_\infty)^{1/2} \\ &= \left[\left(\max_j \sum_i f_{ij}^{rs} \right) \left(\max_i \sum_j f_{ij}^{rs} \right) \right]^{1/2}, \end{aligned}$$

since all matrix elements are positive. It can be verified by calculation that

$$\sum_{j=1}^\infty f_{1j}^{11} < .759,$$

and that this row sum is larger than any other bound that can be obtained for small i (say $i < 20$). Lemma 3 shows that this value certainly cannot be exceeded for any

larger i . (The factors a_i^{rs} and b_i^{rs} in Lemma 2 are exponentially small in i and present no difficulties.) \square

Computations show that the largest singular value σ_{\max} always belongs to F^{11} . A block Lanczos code written by Underwood [30] was used to compute this value for N ranging from 1 to 2047. The results, together with the bound from the proof of Theorem 4, are shown in Fig. 2. It should be noted that the smallest row sum of the matrices F^{rs} can be used to obtain a lower bound on σ_{\max} as well. Calculations indicate that $\sigma_{\max} > .7$ as N tends to infinity.

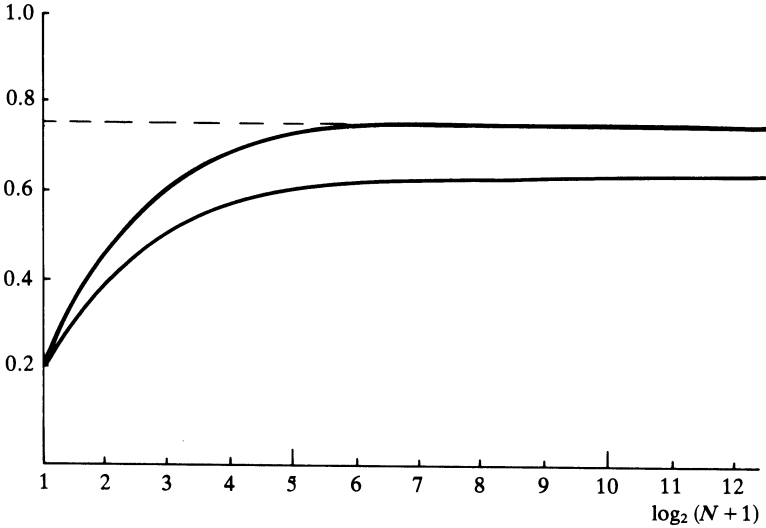


FIG. 2. The largest singular value as a function of $\log_2(N+1)$ (below), compared with the corresponding Gerschgorin bound (above).

The next lemma shows that the singular values σ_i cluster at zero.

LEMMA 4. The following bounds on the sum of the singular values $\{\sigma_i\}$ hold:

$$\sum_{i=1}^{N/2} \sigma_i < \ln N \quad \text{if } \sigma_i \text{ belongs to } F^{11} \text{ or } F^{22},$$

$$\sum_{i=1}^{N/2} \sigma_i^2 < \ln N \quad \text{if } \sigma_i \text{ belongs to } F^{12} \text{ or } F^{21}.$$

Proof. Consider the matrix F^{11} . Since F^{11} is symmetric, it is sufficient to look at its trace.

$$\sum_{i=1}^N F_{ii}^{11} = \sum_{i=1}^N \frac{2}{\pi} \frac{1}{i_1} \left(\frac{a_i^{11}}{b_i^{11}} \right)^2 \leq \frac{1}{\pi} \left(\gamma + \ln N + \delta + O\left(\frac{1}{N}\right) \right)$$

where γ is Euler's constant, $\gamma = .5772 \dots$ and δ is the contribution from the small term a_i^{11}/b_i^{11} . Letting $N \rightarrow \infty$, this shows that the constant in front of the $\ln N$ term in the lemma (taken equal to 1 there) tends to $1/\pi$ as N becomes large. A similar argument gives the same result for F^{22} . It is an obvious conjecture that this result is true also for F^{12} , but since it is of little importance in this context a weaker statement is given. This can be proved by considering $\sum_{ij} (F_{ij}^{12})^2$ (the Frobenius norm of F^{12}). \square

We now conclude this section with two theorems describing the rate of convergence of the conjugate gradient iteration proposed in § 3.

THEOREM 5. *If the conjugate gradient algorithm is used to solve the linear system $T_r x = y$ with the splitting $T_r = \tilde{T}_r - (\tilde{T}_r - T_r)$, then the initial error will be reduced by a factor ε after at most*

$$k = \ln \left(\frac{2}{\varepsilon} \right)$$

iterations.

Proof. Let $\mu_1 \leq \mu_2 \leq \dots \leq \mu_N$ be the eigenvalues of $\tilde{T}_r^{-1} T_r$. It is well known from the theory of the conjugate method [19] that

$$\varepsilon \leq \frac{1}{T_k} \left(\frac{\mu_N + \mu_1}{\mu_N - \mu_1} \right)$$

where T_k is the k th Chebyshev polynomial of the first kind. $T_k(x) = \cosh(k \cosh^{-1} x)$ for $x > 1$. Therefore

$$k \leq \frac{\cosh^{-1}(1/\varepsilon)}{\cosh^{-1} \left(\frac{\mu_N + \mu_1}{\mu_N - \mu_1} \right)}$$

Using $\cosh^{-1}(1/\varepsilon) < \ln(2/\varepsilon)$, $\mu_1 > 1 - .8^2 = .36$, and $\cosh^{-1}((1 + .36)/(1 - .36)) > 1$ gives the desired result. \square

This theorem establishes convergence to any prescribed accuracy in a constant number of iterations independent of N . Since each iteration takes $O(N^2)$ arithmetic operations, the description of an $O(N^2 \log N)$ algorithm for the first biharmonic problem using (16) is complete. If the accuracy is required to increase with increasing N as N^{-p} for a fixed p , then $O(\log N)$ iterations are required and the overall asymptotic operation count remains unchanged. (In order to be consistent with a decreasing discretization error, p should be 2.)

However, under this assumption the use of an $O(N^2)$ Poisson solver will not make the overall algorithm any faster if the solution on the final grid is computed directly. In order to have an $O(N^2)$ method, it is necessary to compute the solution on a sequence of grids, reducing the error by a fixed amount on each grid. (The total work on all the coarser grids will only be $O(N^2)$.)

For practical computations ($N \leq 2047$), the use of the computed spectral radius $\sigma_{\max} = .6343$ for $N = 2047$ (see Fig. 2) strengthens the above theorem to

$$k \leq \frac{1}{2} \ln \left(\frac{2}{\varepsilon} \right).$$

As an illustration, with $\varepsilon = 10^{-10}$ this estimate gives $k \leq 12$.

The above theorems show that the conjugate gradient iteration converges at a very fast linear rate. The next theorem complements this by showing that asymptotically the rate of convergence is in fact superlinear.

Recall that a sequence $\{e_k\}_{k=0}^\infty$ converges R -superlinearly to zero if and only if $\lim_{k \rightarrow \infty} \sup \|e_k\|^{1/k} = 0$. An excellent reference discussing the convergence of iterative processes is Ortega and Rheinboldt [22].

THEOREM 6. *The conjugate gradient method defined in Theorem 5 has an R -superlinear rate of convergence.*

Proof. Using the optimality property of the conjugate gradient iteration,

$$\|e_k\| \equiv (c_k)^k \|e_0\| \leq \max_{\mu \in \{\mu_i\}_{i=1}^N} \prod_{j=1}^k \left| \frac{\mu_j - \mu}{\mu_j} \right| \|e_0\|$$

where $\|e_k\|$ is the error in the appropriate norm at iteration k . Let the set $\{\mu_i\}_{i=1}^N$ be ordered such that $\mu_i \leq \mu_{i+1}$ for all i . Then

$$\begin{aligned} \|e_k\| &\leq \max_{\mu \in \{\mu_i\}_{i=k+1}^N} \prod_{j=1}^k \left| \frac{\mu_j - \mu}{\mu_j} \right| \|e_0\| \\ &\leq \max_{\sigma \in \{\sigma_j\}_{j=k+1}^N} \prod_{j=1}^k \frac{\sigma_j^2 - \sigma^2}{1 - \sigma_j^2} \|e_0\| \\ &\leq \prod_{j=1}^k \frac{\sigma_j^2}{1 - \sigma_j^2} \|e_0\|. \end{aligned}$$

Using the arithmetic-geometric mean inequality, Lemma 4 and the fact that $\sigma_j < 1$ for all j gives

$$\|e_k\| \leq \left(\frac{1}{k} \sum_{j=1}^k \frac{\sigma_j^2}{1 - \sigma_j^2} \right)^k \|e_0\| \leq \left(\frac{1}{k} \frac{\ln N}{1 - \sigma_1^2} \right)^k \|e_0\|.$$

This inequality shows that the constant

$$c_k \leq \frac{1}{k} \frac{\ln N}{1 - \sigma_1^2}$$

tends to zero as k increases for fixed N .

However, since the concept of R -superlinear convergence is most meaningful in the case of an infinite number of iterations and the conjugate gradient method has finite termination on finite-dimensional problems, consider the limiting case as $N \rightarrow \infty$. Lemma 4 implies that $\lim_{k \rightarrow \infty} \sigma_k = 0$, and therefore

$$\lim_{k \rightarrow \infty} c_k = \lim_{k \rightarrow \infty} \left(\frac{1}{k} \sum_{j=1}^k \frac{\sigma_j^2}{1 - \sigma_j^2} \right) = 0. \quad \square$$

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