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Non-Self-Adjoint Linear Systems II

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Iterative Methods for Cyclically Reduced
Non-Self-Adjoint Linear Systems II

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Abstract

We perform an analytic and experimental study of line iterative methods for solving linear systems arising from **finite** difference **discretizations** of non-self-adjoint elliptic partial differential equations on two-dimensional domains. The methods consist of performing one step of cyclic reduction, followed by solution of the resulting reduced system by line relaxation. We augment previous analyses of one-line methods, and we derive a new convergence analysis for two-line methods, showing that both classes of methods are highly effective for solving the **convection-diffusion** equation. In addition, we compare the experimental **performance** of several variants of these methods, and we show that the methods can be implemented **efficiently** on parallel architectures.

Abbreviated Title. Iterative Methods for Reduced Systems.

Key words: Linear systems, reduced system, iterative methods, convection-diffusion, non-self-adjoint.

AMS(MOS) subject classification. Primary: **65F10, 65N20**. Secondary: **15A06**.

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1. Introduction.

We consider iterative methods for solving linear systems of the type that arise from two-cyclic **discretizations** of t -dimensional elliptic partial differential equations. Such systems **can** be ordered using a *red-black ordering* so that they have the form

$$(1.1) \quad \begin{pmatrix} \mathbf{D} & \mathbf{C} \\ \mathbf{E} & \mathbf{F} \end{pmatrix} \begin{pmatrix} \mathbf{u}^{(r)} \\ \mathbf{u}^{(b)} \end{pmatrix} = \begin{pmatrix} \mathbf{v}^{(r)} \\ \mathbf{v}^{(b)} \end{pmatrix}$$

where \mathbf{D} and \mathbf{F} are diagonal matrices. If block elimination is used to decouple the “red” points $\mathbf{u}^{(r)}$ from the “black” points $\mathbf{u}^{(b)}$, the result is a *reduced system*

$$(1.2) \quad [\mathbf{F} - \mathbf{E}\mathbf{D}^{-1}\mathbf{C}]\mathbf{u}^{(b)} = \mathbf{v}^{(b)} - \mathbf{E}\mathbf{D}^{-1}\mathbf{v}^{(r)}.$$

Let

$$(1.3) \quad \mathbf{S} = \mathbf{F} - \mathbf{E}\mathbf{D}^{-1}\mathbf{C}, \quad \mathbf{s} = \mathbf{v}^{(b)} - \mathbf{E}\mathbf{D}^{-1}\mathbf{v}^{(r)}.$$

In [4], we showed that the **coefficient** matrix \mathbf{S} is also **sparse**, and we analyzed a class of iterative methods for solving (1.2) when (1.1) comes from a **finite-difference** discretization of the constant coefficient **convection-diffusion** equation

$$(1.4) \quad \mathcal{A}\mathbf{u} = -\Delta\mathbf{u} + \sigma\mathbf{u}_x + \tau\mathbf{u}_y = \mathbf{f}$$

with Dirichlet boundary conditions. In particular, we showed that although \mathbf{S} is typically nonsymmetric, it can be symmetrized in a wide variety of **circumstances**. The symmetrized form was used to analyze the convergence properties of a splitting operator based on a block Jacobi splitting of \mathbf{S} , using a one-line *ordering* of the underlying grid.

In this paper, we **refine** and augment the analysis of [4]. We show that if (1.1) is derived from the convection-diffusion equation (1.4), then the reduced system is itself a discretization of the differential equation. We consider a variety of orderings of the rows and columns of \mathbf{S} and examine their effects on the convergence of iterative methods for solving (1.2), and on implementation. In **particular**, we present several variants of the one-line ordering of [4] based on red-black and **toroidal** groupings of unknowns. In addition, we present an analysis of **two-line** ordering strategies for solving (1.2); such orderings have been studied for **self-adjoint** problems in [6], [12]. In all of these **cases**, the reduced matrices have block Property A so that Young’s analysis of iterative methods [18] is applicable. We **use** this **analysis** to determine the convergence **properties** of block Jacobi, Gauss-Seidel and successive **overrelaxation** (SOR) methods for solving the discrete convection-diffusion equation, in terms of **discrete** cell Reynolds numbers $ah/2$ and $\tau h/2$. In addition, we present the results of numerical experiments showing some effects of ordering strategies not revealed by the analysis. Together, the analytic and numerical results show **that** the two types of orderings are **very effective** for solving (1.4), with the two-line orderings somewhat more **effective** than the one-line **orderings**. The **variants** of the methods based on red-black orderings of the reduced system are typically **slightly** slower (in terms of iteration counts), but they **can be** implemented more **efficiently** on parallel architectures.

An outline of the paper is as follows. In §2, we describe two discretization schemes for (1.4), and we present an analysis of the truncation error associated with taking the reduced system as an approximation of (1.4). In §3, we present several variants of the **one-line** ordering for the unknowns of (1.2), and we show how the results of [4] are used to derive a convergence analysis of all the associated one-line iterative methods when the linear system comes from (1.4). In §4, we present **the two-line** orderings and the convergence analysis of the corresponding two-line iterative methods applied to (1.2). In §5, we outline an **analysis** due to Parter [12] and Parter and **Steuerwalt** [14] that complements our results in the **limiting** case $h \rightarrow 0$. In §6, we describe numerical experiments that **confirm** and supplement the convergence analysis, including tests in which the block iterative methods, with various orderings, are used to solve a set of nonsymmetric problems derived from (1.4). Finally, in §7 we draw some conclusions.

2. The convection-diffusion equation and the reduced system.

Consider the two-dimensional **convection-diffusion** equation (1.4), posed on the unit square $\Omega \in (0, 1) \times (0, 1)$ with Dirichlet boundary conditions $u = g$ on $\partial\Omega$. **Discretization** by a five-point **finite** difference operator leads to a linear system

$$A \cdot u = v$$

where u now denotes a vector in a **finite** dimensional space. We discretize on a uniform $n \times n$ grid using standard second order differences for the Laplacian [17], [18], and either centered or upwind differences for the **first** derivatives. With u ordered lexicographically in the natural ordering as $(u_{1,1}, u_{2,1}, \dots, u_{n,n})^T$, the **coefficient** matrix has the form

$$(2.1) \quad A = \text{tri} [A_{j,j-1}, A_{jj}, A_{j,j+1}].$$

Here, $\text{tri} [X_{j,j-1}, X_{jj}, X_{j,j+1}]$ is the (block) tridiagonal matrix whose j 'th row contains $X_{j,j-1}$, X_{jj} and $X_{j,j+1}$ on its subdiagonal, diagonal and superdiagonal, respectively. The subdiagonal of the first row and the **superdiagonal** of the last row are not **defined**. The *subscripts will be omitted when there is no ambiguity. The entries of (2.1) are

$$A_{j,j-1} = bI, \quad A_{jj} = \text{tri} [c, a, d], \quad A_{j,j+1} = eI,$$

where I is the identity matrix, a, b, c, d and e depend on the discretization, and all blocks are of order n . Let $h = 1/(n + 1)$. After scaling by h^2 , the matrix entries are given by

$$\begin{aligned} a &= 4, & b &= -(1 + \delta), & c &= -(1 + \gamma), \\ d &= -(1 - \gamma), & e &= -(1 - \delta), \end{aligned}$$

for the centered **difference** scheme, where $\gamma = \sigma h/2$ and $\delta = \tau h/2$; and

$$\begin{aligned} a &= 4 + 2(\gamma + \delta), & b &= -(1 + 2\delta), & c &= -(1 + 2\gamma), \\ d &= -1, & e &= -1, \end{aligned}$$

for the **upwind** scheme. At the (i, j) grid point, the right hand side **satisfies** $v_{ij} = h^2 f_{ij}$ where $f_{ij} \equiv f(ih, jh)$.

In [4], we showed that the reduced matrix S is a **skewed** nine-point operator. At **all** grid points except those bordering $\partial\Omega$, the computational molecule has the form (after scaling by a) given in Fig. 2.1. For grid points next to $\partial\Omega$, the diagonal entries of S (center point of the computational molecule) are **different**. These values are

(2.2)
$$\begin{aligned} a^2 - 2be - cd & \text{ for points with one horizontal and two vertical neighbors} \\ & \text{in the original grid} \\ a^2 - be - 2cd & \text{ for points with one vertical and two horizontal neighbors} \\ a^2 - be - cd & \text{ for points with just two neighbors.} \end{aligned}$$

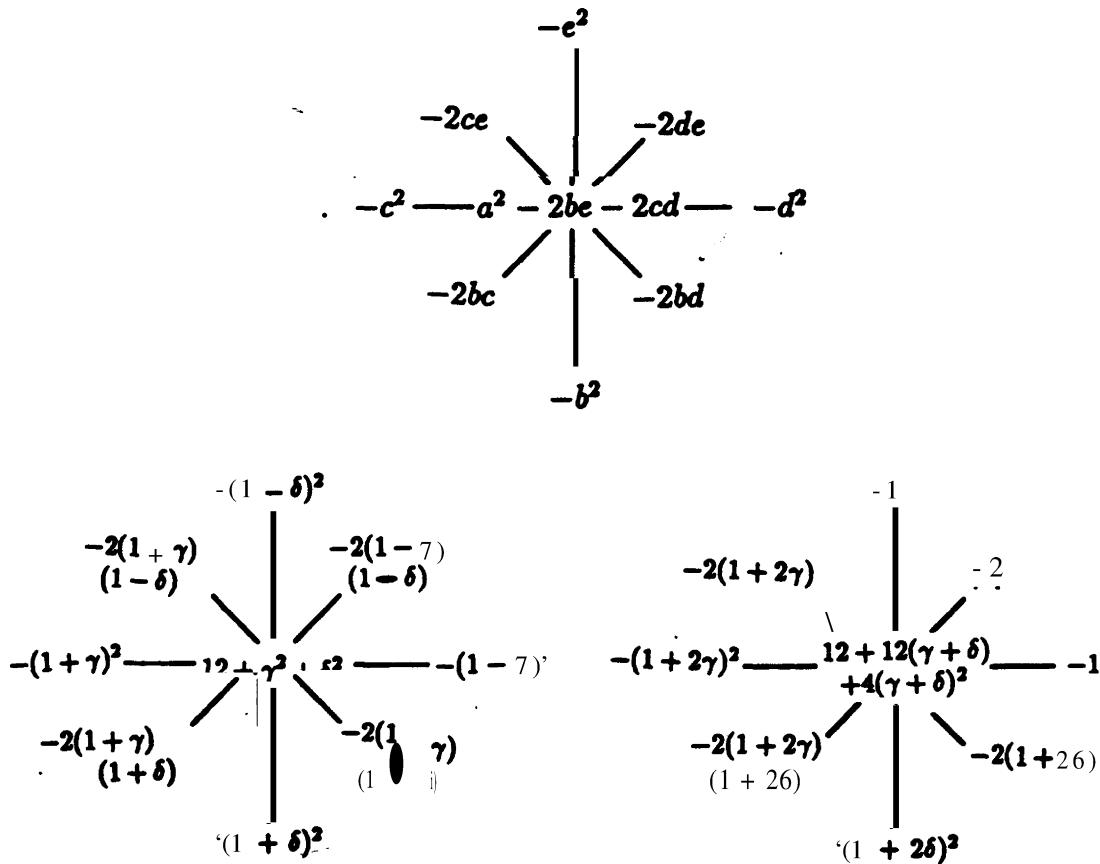


Fig. 2.1: Computational **molecules for** the reduced system. Top: general case. Bottom left: centered **differences**. Bottom right: upwind **differences**.

Suppose centered **differences are used** to **discretize** the **first** derivative terms. At the (i, j) **grid point**, the discrete-operator, satisfies $\frac{1}{h^2}[\mathbf{A}\mathbf{u}]_{ij} = [\mathbf{A}\mathbf{u}]_{ij} + O(h^2)$, i.e. the truncation error of the discretization is of order h^2 . The following result shows that the

reduced system (1.2) can also be viewed as a discretization of (1.4) with truncation error of order h^2 . When (1.2) arises from the centered difference discretization of (1.4), let $\tilde{\mathbf{S}}$ and $\tilde{\mathbf{s}}$ denote the reduced matrix and right hand side resulting from multiplying the reduced system by a ($= 4$).

THEOREM 1. For $2 \leq i, j \leq n - 1$, the reduced operator $\tilde{\mathbf{S}}$ satisfies

$$\frac{1}{8h^2}[\tilde{\mathbf{S}}\mathbf{u}]_{ij} = -\left[\left(1 + \frac{\sigma h^2}{8}\right)u_{xx} + \left(1 + \frac{\tau h^2}{8}\right)u_{yy}\right] + \sigma u_x + \tau u_y + O(h^2),$$

and the reduced right hand side $\tilde{\mathbf{s}}$ satisfies

$$\frac{1}{8h^2}\tilde{s}_{ij} = f_{ij} + O(h^2).$$

Proof. The proof follows directly by taking the first five terms of the Taylor series for each of the quantities $u_{i\pm 2,j}$, $u_{i,j\pm 2}$ and $u_{i\pm 1,j\pm 1}$, expanded about u_{ij} . Multiplying each entry of $\tilde{\mathbf{S}}$ (e.g. from Fig. 2.1) by the appropriate expanded value of u and summing the coefficients for each partial derivative gives

$$\begin{aligned} [\tilde{\mathbf{S}}\mathbf{u}]_{ij} &= 8h^2 \left[\sigma u_x + \tau u_y - \left(1 + \frac{\sigma h^2}{8}\right)u_{xx} - \left(1 + \frac{\tau h^2}{8}\right)u_{yy} - \frac{\sigma\tau}{4}h^2 u_{xy} \right. \\ &\quad + \frac{5}{12}\sigma h^2 u_{xxx} + \frac{1}{4}\tau h^2 u_{xxy} + \frac{1}{4}\sigma h^2 u_{zyy} + \frac{5}{12}\tau h^2 u_{yyy} \\ &\quad \left. - \frac{5}{24}h^2 u_{xxxx} - \frac{1}{4}h^2 u_{xxyy} - \frac{5}{24}h^2 u_{yyyy} + O(h^4) \right]. \end{aligned}$$

The reduced right hand side is given by

$$\tilde{s}_{ij} = 4v_{ij} + (1 + \delta)v_{i,j-1} + (1 + \gamma)v_{i-1,j} + (1 - \gamma)v_{i+1,j} + (1 - \delta)v_{i,j+1}.$$

Using the fact that $v_{ij} = h^2 f_{ij}$ for all (i, j) and expanding $f_{i,j\pm 1}$ and $f_{i\pm 1,j}$ in Taylor series about f_{ij} gives

$$\tilde{s}_{ij} = 8h^2 f_{ij} - h^4(-\Delta f + oft + \tau f_y) + O(h^5). \quad \square$$

The expression for $[\tilde{\mathbf{S}}\mathbf{u}]_{ij}$ in this proof was computed by hand and checked using MACSYMA [9]. The perturbation of the Laplacian (which is also of order h^2) can be thought of as an addition of artificial viscosity, see [16]. A similar analysis shows that the reduced system for the upwind scheme approximates (1.4) with truncation error $O(h)$.

In the following, we use the symbols \mathbf{S} and \mathbf{s} to represent the reduced matrix and right hand side, respectively, after scaling by the diagonal entry a . Our analysis of iterative methods for solving the reduced system (1.2) is based on the fact that in some circumstances, \mathbf{S} can be symmetrized by a real diagonal similarity transformation.

THEOREM 2. There exists a real diagonal matrix \mathbf{Q} such that $\mathbf{Q}^{-1}\mathbf{S}\mathbf{Q}$ is symmetric if the product $bcde$ is positive.

See [4] for a proof. For the centered difference scheme, $be = 1 - \delta^2$ and $cd = 1 - \gamma^2$, so that \mathbf{S} is symmetrizable if both $|\gamma| < 1$ and $|\delta| < 1$ or if both $|\gamma| > 1$ and $|\delta| > 1$. For the

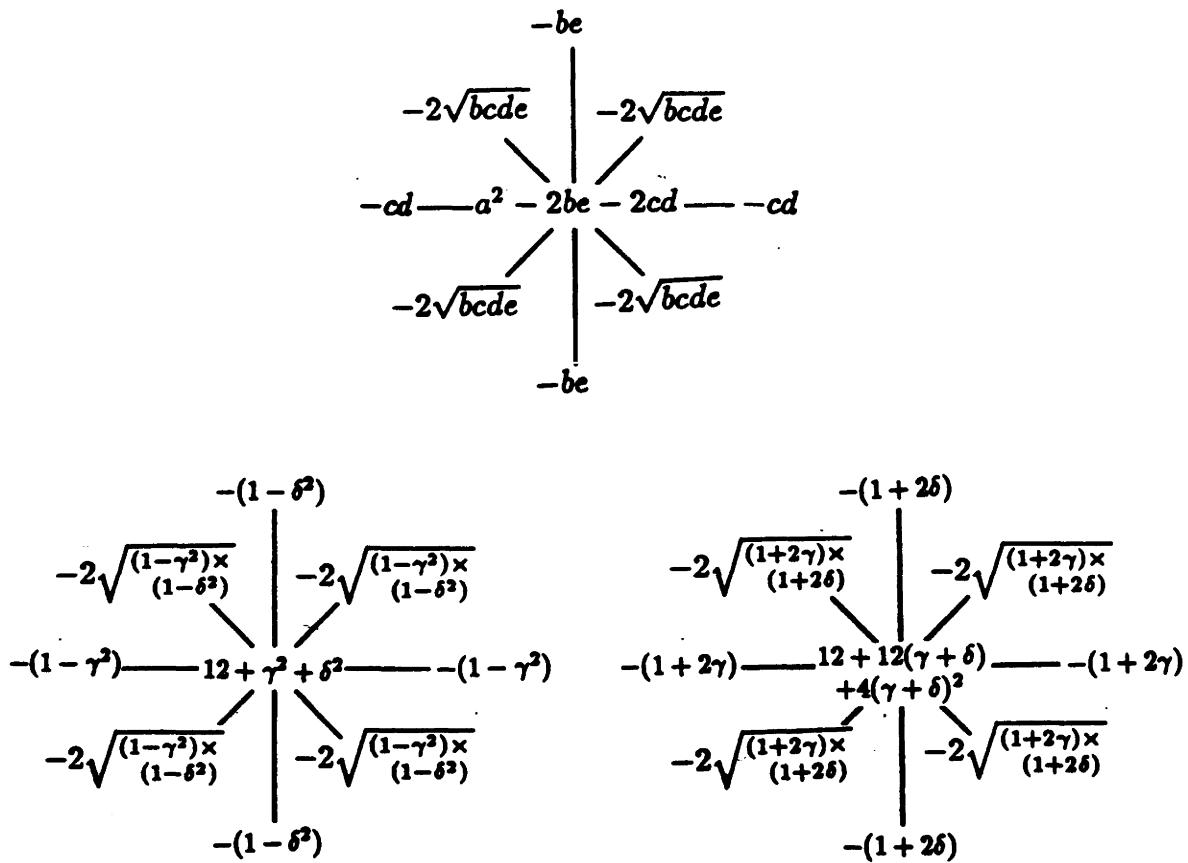


Fig. 2.2: Computational molecules for the symmetrized reduced system. Top: general case. Bottom left: centered differences. Bottom right: upwind differences.

upwind scheme, S is **symmetrizable** for all nonnegative γ and δ . Computational molecules for the symmetrized matrix are shown in Fig. 2.2.

3. One-he orderings.

The performance of iterative methods for solving (1.2) depends on the ordering of the **underlying grid**. In this **section**, we describe and analyze several **one-line** orderings, in which **grid points are grouped by diagonal lines** oriented at a 45° angle with the horizontal and vertical **axes**. For the **purpose** of discussion, we **fix** the orientation to be along the NW-SE **direction**. We consider four orderings.

In the **natural one-line ordering**, the $n - 1$ diagonal **lines** are numbered starting **from** one corner (e.g. the SW) **from** 1 to $n - 1$, and individual points are numbered **from** bottom to top **along** the lines. An example for $n = 7$ is shown in the left side of Fig. 3.1 where the line indices **are shown** outside $\partial\Omega$. The **corresponding** matrix S is block **tridiagonal**. In the **red-black one-line ordering**, the **lines** with odd indices **from** the natural ordering are ordered **first**, followed by **those** with even indices. The individual grid points are renumbered **to**

be consistent with this reordering. An example for $n = 7$ is shown in the right side of Fig. 3.1. Here, the reduced matrix has the form of the coefficient matrix of (1.1), where the block diagonal matrices D and F consist of uncoupled tridiagonal blocks.⁴

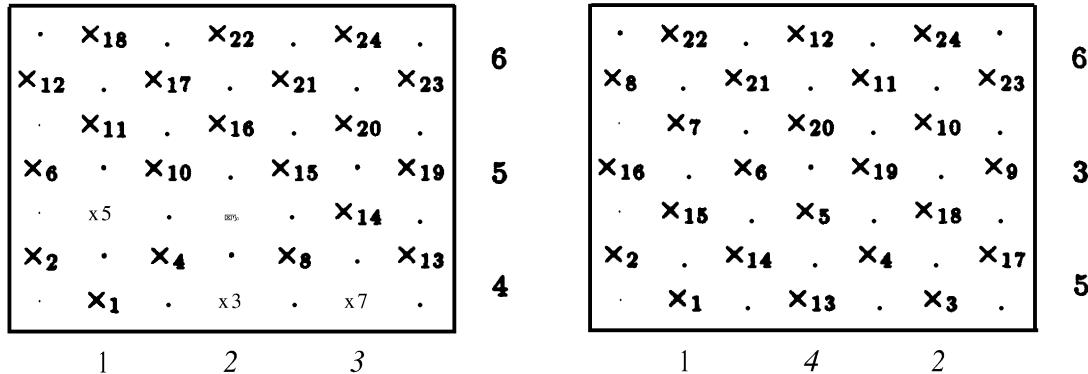


Fig. 3.1: The reduced grid derived from a 7×7 grid, with natural one-line (left) and red-black one-line (right) orderings.

Note that the individual lines in the reduced grid, and therefore the associated **tridiagonal** matrices, vary in size. For the natural ordering, the lines have sizes

$$\begin{cases} 2, 4, \dots, n-1, n-1, \dots, 4, 2 & \text{for odd } n \\ 2, 4, \dots, n-2, n, n-2, \dots, 4, 2 & \text{for even } n. \end{cases}$$

The other two orderings are **defined** so that lines of less than maximal size are paired up to form sets of **fixed** size. This will be of use on parallel architectures (see §6). In the **torus one-line** ordering, each line of less than maximal size from one corner of the grid is followed by the line from the **opposite** corner that would be obtained by continuing the grid periodically; these pairs of lines then are organized as in the natural ordering. For example, for odd n , the first four lines are the one in the SW corner containing **2** mesh points, followed by the line closest to the NE corner containing $n - 3$ points, the line of size **4** in the SW corner and then the line of size $n - 5$ closest to the NE corner. The ordering for $n = 7$ is shown in the left side of Fig. 3.2. Thus, the reduced grid can be grouped together into **[n/2] sets** consisting of either one or two lines, each containing a total of $n - 1$ mesh **points**. For even n , the analogue produces $n/2$ sets of points, each of size **n**.

We **define** the **fourth** ordering in terms of these **fixed** sized sets. Suppose first that they are listed **consecutively according** to their appearance in the torus ordering. For example, for the grid on the left side of Fig. 3.2, the listing **is**

$$\{1, 2\}, \{3, 4\}, \{5\}, \{6\},$$

where these integers are those outside the domain in the **left** side of Fig. 3.2. Now let this listing be permuted in alternating **fashion**,

$$\{1, 2\}, \{5\}, \{3, 4\}, \{6\},$$

and assign indices of increasing value to these sets. As above, let the grid point indices be assigned so that they are consistent with this ordering of lines. We refer to the result as the *alternating torus one-line* ordering. An example is shown in the right side of Fig. 3.2. This ordering is well-defined for all n , but we will show below that it is most useful when $\lceil n/2 \rceil$ is even, where it corresponds to a red-black ordering.

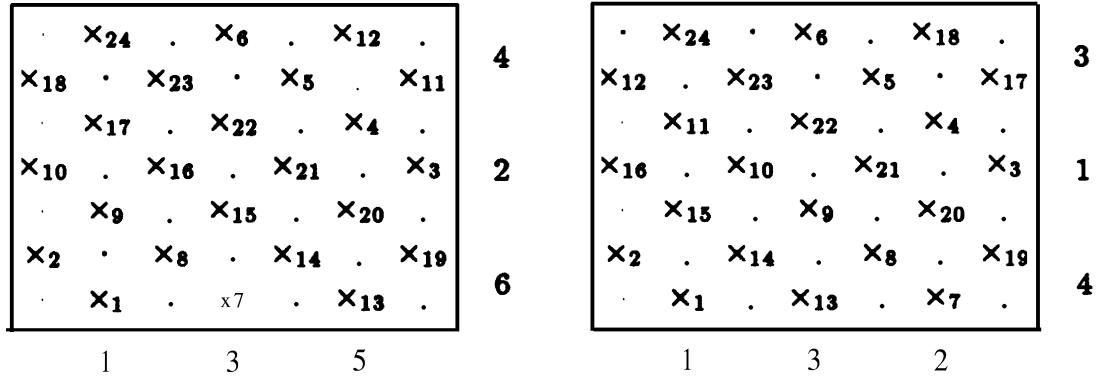


Fig. 3.2: The reduced grid derived from a 7×7 grid, with torus one-line (left) and alternating torus one-line (right) orderings.

For all four orderings, the reduced matrix has the form

$$(3.1) \quad S = D - C,$$

where D is a block diagonal matrix whose individual blocks are tridiagonal matrices. Consider the block Jacobi iterative method

$$u_{k+1}^{(b)} = Bu_k^{(b)} + D^{-1}s,$$

where $B = D^{-1}C$ is the block Jacobi iteration matrix. The standard measure of the effectiveness of this method is the spectral radius $\rho(B)$; the iteration is **convergent** provided $\rho(B) < 1$, and convergence is more rapid if $\rho(B)$ is **closer** to 0 [17] (cf. §6). We have left unspecified the particular ordering determining (3.1); the iteration matrices for the various ordering strategies are all similar to one another, so that $\rho(B)$ is independent of ordering. In [4], we derived bounds on $\rho(B)$ for the version of B arising from the natural ordering. The results are **summarized as** follows; see [4] for proofs.

THEOREM 3. **For the one-line orderings, if** $be > 0$ **and** $cd > 0$, **then**

$$\rho(B) \leq \frac{2(\sqrt{be} + \sqrt{cd})^2}{a^2 - 2(\sqrt{be} + \sqrt{cd})^2 + 4\sqrt{bcde}(1 - \cos(\pi h))}.$$

If $be < 0$, $cd < 0$, **then**

$$\rho(B) \leq \frac{\max(4\sqrt{bcde}, 2\sqrt{bcde} + |be|, 2\sqrt{bcde} + |cd|, |be| + |cd|) + 2(|be| + |cd|)}{a^2 + 2(\sqrt{|be|} - \sqrt{|cd|})^2 + 4\sqrt{bcde}(1 - \cos(\pi h))}.$$

whenever $a^2/2 + (\sqrt{-cd} - \sqrt{-be})^2 - 2\sqrt{bcde} \geq 0$.

COROLLARY 1. For the centered **difference** scheme, if $|\gamma| \leq 1$ and $|\delta| < 1$, then the one-line Jacobi iteration matrices **satisfy**

$$\rho(B) \leq \frac{(\sqrt{1-\gamma^2} + \sqrt{1-\delta^2})^2}{8 - (\sqrt{1-\gamma^2} + \sqrt{1-\delta^2})^2 + 2\sqrt{(1-\gamma^2)(1-\delta^2)(1-\cos(\pi h))}}.$$

If $|\gamma| > 1$, $|\delta| > 1$ and $\sqrt{(\gamma^2 - 1)(\delta^2 - 1)} \leq 4$, then

$$\rho(B) \leq \frac{\frac{1}{2}\mu(\gamma, \delta) + \gamma^2 - 1 + \delta^2 - 1}{8 + (\sqrt{\gamma^2 - 1} - \sqrt{\delta^2 - 1})^2 + 2\sqrt{(\gamma^2 - 1)(\delta^2 - 1)(1 - \cos(\pi h))}},$$

where

$$\mu(\gamma, \delta) = \max(4\sqrt{(\gamma^2 - 1)(\delta^2 - 1)}, 2\sqrt{(\gamma^2 - 1)(\delta^2 - 1)} + \gamma^2 - 1, 2\sqrt{(\gamma^2 - 1)(\delta^2 - 1)} + \delta^2 - 1, \gamma^2 - 1 + \delta^2 - 1).$$

For the upwind **difference** scheme,

$$\rho(B) \leq \frac{(\sqrt{1+2\gamma} + \sqrt{1+2\delta})^2}{2(2 + \gamma + \delta)^2 - (\sqrt{1+2\gamma} + \sqrt{1+2\delta})^2 + 2\sqrt{(1+2\gamma)(1+2\delta)(1-\cos(\pi h))}}.$$

We now show that Young's analysis of relaxation methods also applies to these splittings. Let $C = L + U$, where L and U are strictly lower triangular and upper triangular, and let $\mathcal{L}_\omega = (D - \omega L)^{-1}[(1-\omega)D + \omega U]$ denote the block SOR iteration matrix. Recall the **definition** of block-consistent **orderings** from [18]: a block matrix $M = M_{ij}$, $\{1 \leq i, j \leq m\}$ is block-consistently ordered if the integers $1, \dots, m$ can be partitioned into disjoint sets $\{\mathcal{S}_k\}_{k=1}^t$ such that if $M_{ij} \neq 0$, then $i \in \mathcal{S}_k$ implies $j \in \mathcal{S}_{k-1}$ for $j < i$, and $j \in \mathcal{S}_{k+1}$ for $j > i$. We have the following result:

LEMMA 1. For the natural, red-black and **torus orderings**, the reduced **matrix is block-consistently ordered for all** $n > 0$. For the alternating torus ordering, the reduced **matrix is block-consistently ordered if and only if** $\lceil n/2 \rceil$ is even.

Proof. The **coefficient** matrix for the natural ordering is block tridiagonal; the analysis for this ordering and its red-black analogue is classical, see [18]. In discussing the torus and alternating torus orderings, it will be convenient to refer to the line indices of the natural ordering (i.e. from the left side of Fig. 3.1). Let ψ be a mapping of these indices to those of the **torus** ordering? Then $\mathcal{S}_k = \{\psi(k)\}$ determines a consistent ordering. For the alternating torus ordering, note that its block structure is different than for the other orderings, because pairs of lines are grouped together: using the indices of the natural ordering, lines j and $\lceil n/2 \rceil + j$ are coalesced into one set. If $\lceil n/2 \rceil$ is even, then j and $\lceil n/2 \rceil + j$ have the **same** parity, and the **ordinary** red-black coloring of **lines** determines a

¹ For example, for Figs. 3.1-3.2, $\psi(1)=1, \psi(2)=3, \psi(3)=5$, etc. It is possible to derive a **precise** expression for ψ , but we do not **believe** it adds insight.

red-black coloring of the alternating torus ordering. A consistent ordering is determined by the partitioning

$$\mathcal{S}_1 = \{\{1, \lceil n/2 \rceil + 1\}, \{3, \lceil n/2 \rceil + 3\}, \dots\}, \quad \mathcal{S}_2 = \{\{2, \lceil n/2 \rceil + 2\}, \{4, \lceil n/2 \rceil + 4\}, \dots\}.$$

If $\lceil n/2 \rceil$ is odd, then lines 1 and $\lceil n/2 \rceil + 1$ have the same color, since they comprise one set, but line $\lceil n/2 \rceil$ must share this same color, since (proceeding from the SW corner) alternating lines are assigned opposite colors (see Fig. 3.3). As a result, the alternating torus ordering does not have block Property A, and it therefore cannot be consistently ordered (see [18], §5.4). \square

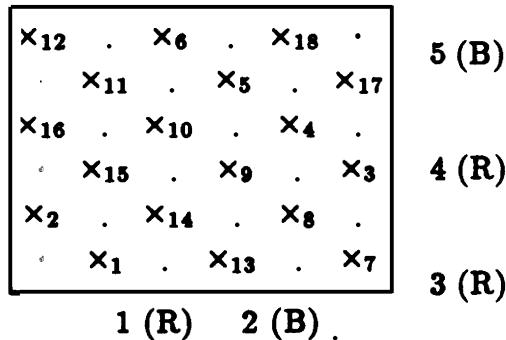


Fig. 3.3: The alternating torus ordering for odd $\lceil n/2 \rceil$. Line indices correspond to the natural one-line ordering. Terms in parentheses indicate that the associated matrix does not have block Property A.

THEOREM 4. The eigenvalues $\{\mu\}$ of B and $\{\lambda\}$ of \mathcal{L}_ω are related by

$$(3.2) \quad (\lambda + w - 1)^2 = \omega^2 \mu^2 \lambda.$$

Moreover, if $p(B) < 1$ and either $be > 0$ and $cd > 0$ holds or $be < 0$ and $cd < 0$ holds, then the choice

$$(3.3) \quad \omega^* = \frac{2}{1 + \sqrt{1 - p(B)^2}}$$

minimizes $p(\mathcal{L}_\omega)$ with respect to w , and $p(\mathcal{L}_{\omega^*}) = \omega^* - 1$.

Proof. The first assertion follows directly from [18], Chapter 14, Theorem 3.4. For the second assertion, it was shown in [4] that if either condition on be and cd holds, then $\hat{\mathbf{D}}$ is a symmetric positive definite M-matrix. Consequently, all eigenvalues of $\hat{\mathbf{B}}$, whence those of B , are real. Therefore, the choice of optimal SOR parameter follows from [18], §5.2 and §14.3. \square

Remarks. When $be > 0$ and $cd > 0$, a sufficient condition to ensure that $p(B) < 1$ is

$$(3.4) \quad a^2 \leq 4(\sqrt{be} + \sqrt{cd})^2,$$

which holds for the two difference schemes. In this case, experiments described in [4] indicate that the bounds of Corollary 1 of Theorem 3 for $|\gamma|, |\delta| < 1$ are good indicators of spectral radii. The bound for $|\gamma|, |\delta| > 1$ of Corollary 1 does not always guarantee that $\rho(B) < 1$. However, experimental evidence and Fourier analysis [4] suggest that the smaller bound

$$\rho(B) \leq \frac{(\sqrt{\gamma^2 - 1} + \sqrt{\delta^2 - 1})^2}{8 + (\sqrt{\gamma^2 - 1} + \sqrt{\delta^2 - 1})^2}$$

applies in this case, and this bound is always less than one. Finally, the results of [5], [10] imply that the Chebyshev semi-iterative method applied to the reduced system, with preconditioning by the block diagonal D , has the same asymptotic convergence behavior as the block SOR method with $w = \omega^*$.

4. Two-line orderings.

An alternative to the ordering strategies of the previous section is to group the points of the reduced grid by **pairs** of horizontal or vertical lines. Such *two-line* orderings also result in matrices that have block Property A. Examples with horizontal lines, for $n = 6$, are shown in Fig. 4.1. The left side of the **figure** shows a *natural two-line ordering*, and the right side shows a *red-black two-line* ordering. In the following, we perform an analysis of two-line orderings for the case of horizontal lines. We use the natural ordering to motivate the analysis; as above, the results also apply to the red-black ordering.

The reduced matrix S for the natural **two-line** ordering **has** block tridiagonal form

$$S = \text{tri} [S_{j,j-1}, S_{j,j}, S_{j,j+1}].$$

Within the line pairs, points are ordered from left to right (as in Fig. 4.1), so that the **submatrices** on the block diagonal are banded. For even n , the block diagonal consists of $n/2$ uncoupled pentadiagonal matrices of order n of the form

$$S_{jj} = \begin{pmatrix} * & -2b d & -d^2 & & & & \\ -2ce & * & -2de & -d^2 & & & \\ -c^2 & -2bc & * & -2b d & -d^2 & & \\ & c^2 & -2ce & * & -2d e & -d^2 & \\ & & & & \ddots & & \\ & & & & & & \end{pmatrix},$$

$1 \leq j \leq n/2$. Here, “*” is defined as in the center point of Fig. 2.1, or by (2.2) for points next to $\partial\Omega$. The off-diagonal blocks have the irregular tridiagonal form

$$S_{j,j-1} = - \begin{pmatrix} b^2 & & & & \\ 2bc & b^2 & 2bd & & \\ & b^2 & & & \\ & 2bc & b^2 & 2bd & \\ & & b^2 & & \end{pmatrix}, \quad S_{j,j+1} = - \begin{pmatrix} e^2 & 2de & & & \\ e^2 & & & & \\ 2ce & e^2 & 2de & & \\ e^2 & & & & \\ 2ce & e^2 & 2de & & \\ & & & & \end{pmatrix}.$$

For odd n , the last ($\lceil n/2 \rceil$ 'th) row and column have slightly different form, in which the last diagonal block is the tridiagonal matrix of order $\lfloor n/2 \rfloor$,

$$(4.1) \quad \text{tri} [-c^2 *, -d^2],$$

and the neighboring off-diagonal blocks are adjusted in an analogous manner.

Let D now denote the block diagonal matrix **defined** by $D_j = S_{jj}$, and let $S = D - C$ denote the two-line Jacobi splitting. Consider the two-line Jacobi iteration

$$u_{k+1}^{(b)} = Bu_k^{(b)} + D^{-1}s,$$

for solving (1.2), where $B = D^{-1}C$. Convergence again depends on $\rho(B)$. Let $\hat{S} = Q^{-1}SQ$ represent the symmetrized reduced matrix when it exists, and let $\hat{D} = Q^{-1}DQ$ and $\hat{C} = Q^{-1}CQ$. We first bound $\rho(B)$ in the case where $be > 0$ and $cd > 0$, i.e. for the centered difference scheme when $|\gamma| < 1$ and $|\delta| < 1$, and for the upwind scheme. The analysis essentially consists of the following two results, which bound the minimum eigenvalue of \hat{D} and maximum eigenvalue of \hat{C} . These will then be combined to bound $p(B)$.

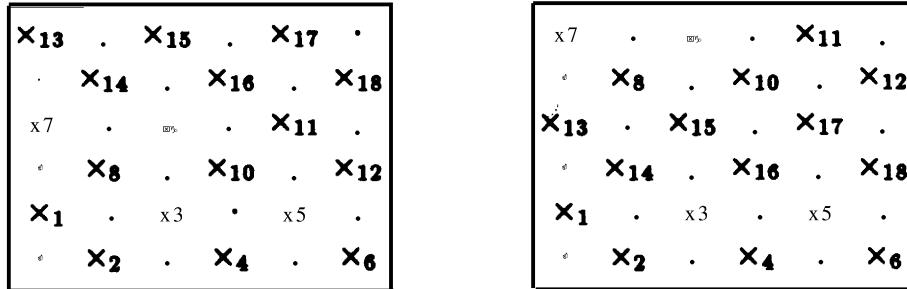


Fig. 4.1: The reduced grid derived from a 6×6 grid, with natural two-line (left) and red-black two-line (right) orderings.

LEMMA 2. When $be > 0$ and $cd > 0$, the minimum eigenvalue of the symmetrized two-line block diagonal matrix \hat{D} is bounded below by

$$a^2 - 2(\sqrt{cd} + \sqrt{be})^2 - 2cd + 4\sqrt{bcde}(1 - \cos \pi h) + 4cd(1 - \cos^2 \pi h).$$

Proof. Examination of Figs. 2.2 and 4.1 reveals that all of the matrices on the block diagonal of \hat{D} , except the first and last, are identical **pentadiagonal** matrices of order n . They have the form

$$(4.2) \quad P \equiv \begin{pmatrix} * & -2\sqrt{bcde} & -cd & & & \\ -2\sqrt{bcde} & * & -2\sqrt{bcde} & -cd & & \\ & -cd & * & -2\sqrt{bcde} & -cd & \\ & -2\sqrt{bcde} & -cd & * & -2\sqrt{bcde} & -cd \\ & & -2\sqrt{bcde} & -cd & * & -2\sqrt{bcde} \\ & & & & -cd & -cd \end{pmatrix},$$

where “*” equals $a^2 - 2be - 2cd$ except in the **first** and last entry, where it is $a^2 - 2be - cd$. If \mathbf{P}' denotes either the **first** block, or for even n , the last block of \hat{D} , then by (2.2), we have $\mathbf{P}' \geq P$ with inequality only on the diagonal. Hence, $\lambda_{\min}(\mathbf{P}') \geq \lambda_{\min}(P)$. A straightforward argument also shows that for all small h , the minimum eigenvalue of \hat{D} does not correspond to an eigenvalue of the tridiagonal matrix (4.1). Hence, it suffices to bound $\lambda_{\min}(P)$ **below**.

For this, let \mathbf{T}_n denote the tridiagonal matrix $\text{tri} [1, 0, 1]$ of order n . Then \mathbf{T}_n^2 is a **pentadiagonal** matrix with 0's on the **first** subdiagonal and **superdiagonal**, 1's on the second subdiagonal and superdiagonal, and 2's in all diagonal entries except the first and last, where the values are 1. Then we have

$$(4.3) \quad P = (a^2 - 2be) \mathbf{I}_n - 2\sqrt{bcde} \mathbf{T}_n - cd \mathbf{T}_n^2,$$

where \mathbf{I}_n is the identity matrix of order n . But the **eigenvalues** of \mathbf{T}_n are $\{2 \cos(j\pi h)\}_{j=1}^n$, so that those of P are $\{a^2 - 2be - 4\sqrt{bcde} \cos(j\pi h) - 4cd \cos^2(j\pi h)\}_{j=1}^n$. The minimum corresponds to the choice $j = 1$. \square

LEMMA 3. **The maximum eigenvalue** of the symmetrized two-line block off-diagonal **matrix** $\hat{\mathbf{C}}$ is bounded by

$$2|be| \cos 2\pi h + 4\sqrt{bcde} \cos \pi h + o(h^2).$$

Proof. Assume n is even; modifications to the argument for odd n are straightforward. Let \mathbf{R} denote the block tridiagonal matrix $\text{tri} [\mathbf{R}, 0, \mathbf{R}]$, with $m = n/2$ block rows, where $\mathbf{R} = be \mathbf{I}_m$. Let \mathbf{V} denote the block tridiagonal matrix $\text{tri} [\mathbf{V}^T, 0, \mathbf{V}]$, of the same order, where

$$\mathbf{V} = \begin{pmatrix} 0 & v & & & & \\ & 0 & & & & \\ & v & 0 & v & & \\ & & & 0 & & \\ & & & & v & \\ & & & & & \ddots \end{pmatrix}$$

and $v = 2\sqrt{bcde}$. Then $\hat{\mathbf{C}} = \mathbf{R} + \mathbf{Y}$. Since $\hat{\mathbf{C}}$ is symmetric, we have

$$\rho(\hat{\mathbf{C}}) = \|\hat{\mathbf{C}}\|_2 \leq \|\mathbf{R}\|_2 + \|\mathbf{Y}\|_2.$$

To bound $\|\mathbf{R}\|_2$, note that \mathbf{R} is **similar** to the block diagonal matrix

$$\text{be diag} \{\mathbf{T}_m, \dots, \mathbf{T}_m\}$$

with n block rows, so that its **eigenvalues** are $\{2be \cos \frac{j\pi}{m+1}\}_{j=1}^m$. Hence

$$(4.4) \quad \|\mathbf{R}\|_2 = \rho(\mathbf{R}) = 2|be| \cos \left(\frac{2\pi h}{1+h} \right).$$

For $\|\mathcal{V}\|_2$, we have

$$\|\mathcal{V}\|_2 = \|\mathcal{V}^T \mathcal{V}\|_2^{1/2} = \|\mathcal{V}^2\|_2^{1/2}.$$

\mathcal{V}^2 is the block **pentadiagonal** matrix

$$\begin{pmatrix} \mathbf{V}\mathbf{V}^T & 0 & \mathbf{V}^2 & & & \\ 0 & \mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T & 0 & \mathbf{V}^2 & & \\ (\mathbf{V}^T)^2 & 0 & \mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T & 0 & \mathbf{V}^2 & \\ & & & \ddots & & \\ & & (\mathbf{V}^T)^2 & 0 & \mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T & 0 \\ & & & & (\mathbf{V}^T)^2 & \mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T \\ & & & & & \mathbf{V}^T\mathbf{V} \end{pmatrix}$$

But $\mathbf{V}^2 = 0$, so that in **fact** \mathcal{V}^2 is a block diagonal matrix, and we need only bound the **spectral** radii of $\mathbf{V}\mathbf{V}^T$, $\mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T$ and $\mathbf{V}^T\mathbf{V}$. We have

$$\mathbf{V}\mathbf{V}^T = \begin{pmatrix} \mathbf{v}^2 & 0 & \mathbf{v}^2 & & & \\ 0 & 0 & 0 & 0 & & \\ \mathbf{v}^2 & 0 & 2\mathbf{v}^2 & 0 & \mathbf{v}^2 & \\ & 0 & 0 & 0 & 0 & 0 \\ & & & & \ddots & \\ & & & & \mathbf{v}^2 & 0 & 2\mathbf{v}^2 & 0 \\ & & & & 0 & 0 & 0 & 0 \end{pmatrix}, \quad \mathbf{V}^T\mathbf{V} = \begin{pmatrix} 0 & 0 & 0 & & & \\ 0 & 2\mathbf{v}^2 & 0 & \mathbf{v}^2 & & \\ 0 & 0 & 0 & 0 & 0 & \\ & & & & \ddots & \\ \mathbf{v}^2 & 0 & 2\mathbf{v}^2 & 0 & \mathbf{v}^2 & \\ 0 & 0 & 0 & 0 & 0 & \mathbf{v}^2 \\ \mathbf{v}^2 & 0 & 0 & 0 & \mathbf{v}^2 & 0 \end{pmatrix}$$

Consequently,

$$\mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T = \begin{pmatrix} \mathbf{v}^2 & 0 & \mathbf{v}^2 & & & \\ 0 & 2\mathbf{v}^2 & 0 & \mathbf{v}^2 & & \\ \mathbf{v}^2 & 0 & 2\mathbf{v}^2 & 0 & \mathbf{v}^2 & \\ & & & \ddots & & \\ & & & \mathbf{v}^2 & 0 & 2\mathbf{v}^2 & 0 \\ & & & \mathbf{v}^2 & 0 & 0 & \mathbf{v}^2 \end{pmatrix} = \mathbf{v}^2 \mathbf{T}_n^2.$$

Thus,

$$(4.5) \quad \rho(\mathbf{V}^T\mathbf{V} + \mathbf{V}\mathbf{V}^T) = \mathbf{v}^2 [\rho(\mathbf{T}_n)]^2 = 16 \mathbf{bcde} \cos^2 \pi h.$$

Moreover, by permuting $\mathbf{V}\mathbf{V}^T$ and $\mathbf{V}^T\mathbf{V}$ so that **nonzero** entries lie on a tridiagonal band in the upper left corner, we **find** that

$$(4.6) \quad \rho(\mathbf{V}^T\mathbf{V}) = \rho(\mathbf{V}\mathbf{V}^T) < \mathbf{v}^2 \rho(2\mathbf{I}_m + \mathbf{T}_m) = 16 \mathbf{bcde} \cos^2 \left(\frac{\pi h}{1+h} \right).$$

Bounds (4.5) and (4.6) are essentially the same as $h \rightarrow 0$, which gives the asymptotic result

$$(4.7) \quad \|\mathcal{V}\|_2 \leq 4 \sqrt{\mathbf{bcde}} \cos \pi h + o(h^2).$$

The conclusion then follows from (4.4) and (4.7). \square

The idea used in both these proofs of squaring the tridiagonal matrix \mathbf{T}_n to generate a **pentadiagonal** matrix appears in [13], for analyzing line iterative methods applied to discrete **biharmonic** problems. A simpler argument than the proof of Lemma 3, based on **Gerschgorin's theorem**, gives the weaker bound $\rho(\hat{\mathbf{C}}) \leq 2|be| + 4\sqrt{bcde}$. This bound is close to the result of Lemma 3, but it is less **useful** for asymptotic analysis as $h \rightarrow 0$.

THEOREM 5. *When $be > 0$ and $cd > 0$, the **spectral radius** of the two-line Jacobi iteration **matrix is bounded by***

$$\frac{2be \cos 2\pi h + 4\sqrt{bcde} \cos \pi h}{a^2 - 2(\sqrt{cd} + \sqrt{be})^2 - 2cd + 4\sqrt{bcde}(1 - \cos \pi h) + 4cd(1 - \cos^2 \pi h)} + o(h^2).$$

Proof. Using the similarity transformation $\mathbf{D}^{-1}\mathbf{C} = \mathbf{Q}\hat{\mathbf{D}}^{-1}\hat{\mathbf{C}}\mathbf{Q}^{-1}$, we have

$$\rho(\mathbf{D}^{-1}\mathbf{C}) = \rho(\hat{\mathbf{D}}^{-1}\hat{\mathbf{C}}) \leq \|\hat{\mathbf{D}}^{-1}\|_2 \|\hat{\mathbf{C}}\|_2 = \frac{\rho(\hat{\mathbf{C}})}{\lambda_{\min}(\hat{\mathbf{D}})},$$

where the last **equality** follows from the symmetry of $\hat{\mathbf{D}}$ and $\hat{\mathbf{C}}$. The result then follows immediately from Lemmas 2 and 3. \square

Substitution of particular **values** of a - e gives the **following** bounds for the two difference schemes under consideration.

COROLLARY 2. *For the centered **difference scheme**, if $|\gamma| < 1$ and $|\delta| < 1$, then the **spectral radius** of the two-line block Jacobi iteration **matrix for the reduced system is bounded by***

$$\frac{(1 - \delta^2) \cos 2\pi h + 2\sqrt{(1 - \gamma^2)(1 - \delta^2)} \cos \pi h}{8 - (\sqrt{1 - \gamma^2} + \sqrt{1 - \delta^2})^2 - (1 - \gamma^2) + 2\sqrt{(1 - \gamma^2)(1 - \delta^2)}(1 - \cos \pi h) + 2(1 - \gamma^2)(1 - \cos^2 \pi h)} + o(h^2).$$

For the upwind difference scheme, the **spectral radius is bounded by**

$$\frac{(1 + 2\delta) \cos 2\pi h + 2\sqrt{(1 + 2\gamma)(1 + 2\delta)} \cos \pi h}{2(2 + \gamma + \delta)^2 - (\sqrt{1 + 2\gamma} + \sqrt{1 + 2\delta})^2 - (1 + 2\gamma) + 2\sqrt{(1 + 2\gamma)(1 + 2\delta)}(1 - \cos \pi h) + 2(1 + 2\gamma)(1 - \cos^2 \pi h)} + o(h^2).$$

If (3.4) holds, **then** the bounds of Theorem 5 are **smaller** than those of Theorem 3 for the **one-line orderings**. **Consequently**, the **two-line** bounds of **Corollary 2** are **smaller** than the one-line bound⁸ of Corollary 1.

Now consider the case $be < 0$ and $cd < 0$, which corresponds to the centered difference scheme when $|\gamma| > 1$ and $|\delta| > 1$. To bound $\rho(\mathbf{B})$, we require an **alternative** to Lemma 2. Consider the case of odd n .² Let \mathbf{P} be as in (4.2), where “*” now represents

$$(4.8) \quad \min(a^2 - be - 2cd, a^2 - 2be - cd) \quad (= \min(13 + 2\gamma^2 + \delta^2, 13 + \gamma^2 + 2\delta^2)).$$

² In **this case**, only the **first two terms** of (2.2) occur. For even n , a somewhat weaker bound can be derived by replacing “*” with $a^2 - be - cd$.

For any pentadiagonal matrix \mathbf{P}' on the block diagonal of $\hat{\mathbf{D}}$, the diagonal entries of \mathbf{P}' are greater than those of P , so that $\lambda_{\min}(\mathbf{P}') \geq \lambda_{\min}(P)$. If (4.8) is minimized by $a^2 - be - 2cd$ (i.e. $\gamma^2 \leq \delta^2$), then P satisfies

$$P = (a^2 - be) I_n - 2\sqrt{bcde} T_n - cd T_n^2,$$

(This differs from (4.3) in the coefficient of I_n .) Consequently, all **eigenvalues** of P have the form

$$a^2 - be - 4\sqrt{bcde} \cos \theta - 4cd \cos^2 \theta,$$

for $\theta \in (0, \pi)$. By elementary calculus, we **find** that this expression is minimized at $\theta = \arccos\left(\frac{1}{2}\sqrt{\left|\frac{be}{cd}\right|}\right)$. The minimum value, a^2 , is a **lower** bound for $\lambda_{\min}(\hat{\mathbf{D}})$. If (4.8) is minimized by $a^2 - 2be - cd$ ($\gamma^2 \geq \delta^2$), then

$$P = (a^2 - 2be + cd) I_n - 2\sqrt{bcde} T_n - cd T_n^2.$$

The same argument shows that its minimum eigenvalue is bounded below by $a^2 - be + cd$. As above, these bounds for $\lambda_{\min}(P)$ are **smaller** than the minimum eigenvalue derived from (4.1). Combining these observations with Lemma 3, we have the following result.

THEOREM 6. *For $be < 0$ and $cd < 0$ and even n , the spectral **radius** of the two-line Jacobi iteration **matrix is** bounded by*

$$\begin{cases} \frac{2|be|\cos 2\pi h + 4\sqrt{bcde} \cos \pi h}{a^2} & \text{when (4.8) is minimized by } a^2 - be - 2cd, \\ \frac{2|be|\cos 2\pi h + 4\sqrt{bcde} \cos \pi h}{a^2 - be + cd} & \text{when (4.8) is minimized by } a^2 - 2be - cd. \end{cases}$$

For the centered **difference discretization** when $|\gamma| > 1$ and $|\delta| > 1$, the **bounds** are

$$\begin{cases} \frac{(\delta^2 - 1)\cos 2\pi h + 2\sqrt{(\gamma^2 - 1)(\delta^2 - 1)}\cos \pi h}{8} & \text{for } \gamma^2 < \delta^2, \\ \frac{(\delta^2 - 1)\cos 2\pi h + 2\sqrt{(\gamma^2 - 1)(\delta^2 - 1)}\cos \pi h}{8 + \frac{1}{2}(\delta^2 - \gamma^2)} & \text{for } \gamma^2 \geq \delta^2. \end{cases}$$

As we show in §6, the **bounds** from Theorem 5 and Corollary 2 appear to be tight, whereas the **results** of Theorem 6 are pessimistic.

Finally, the analysis of [4] implies that for both difference schemes, when $be > 0$ and $cd > 0$, the block pentadiagonal matrix $\hat{\mathbf{D}}$ is a symmetric **positive definite** M-matrix. Hence, we have the following result for the two-line SOR iteration matrix \mathcal{L}_ω .

COROLLARY 3. *For the two-line orderings, the **eigenvalues** $\{\mu\}$ of \mathbf{B} and $\{\lambda\}$ of \mathcal{L}_ω are **related** by (3.2). For the two difference schemes under **consideration**, if $be > 0$ and $cd > 0$ holds, then (3.3) **minimizes** $\rho(\mathcal{L}_\omega)$ with respect to ω , und $\rho(\mathcal{L}_{\omega^*}) = \omega^* - 1$.*

5. Asymptotic analysis.

In this section, we outline the results of Parter [12] and Parter and Steuerwalt [14] that reveal asymptotic convergence rates as $h \rightarrow 0$ for **fixed** σ and τ in (1.4). (See also [11].) We emphasize that we are only **filling** in some minor **details**; **all** the analysis is contained in [12], [14]. Assume that S is a matrix **such that** S/h^2 is a discrete approximation to d with truncation error $o(1)$ at **all** mesh points of Ω not next to the boundary, and $O(1)$ at points next to $\partial\Omega$. Let $S = D - C$ be a splitting. The following result is proved in [14]:

THEOREM 7. Suppose the following **conditions** hold for all small h :

- . (PS1) $\rho(D^{-1}C) < 1$.
- (PS2) $\rho(D^{-1}C)$ is an **eigenvalue** of $D^{-1}C$.
- (PS3) $\|C\|_2$ is bounded independent of h .
- (PS4) There is a smooth **function** q satisfying $q(x, y) \geq q_0 > 0$ on $\bar{\Omega}$, such that

$$(5.1) \quad (Cu, v) = (qu, v) + E$$

where in (5.1), q refers to the **vector** of mesh **values**, and $E = h\mathbf{e}_1(u, v) + h^2\mathbf{e}_2(u, v)$ depends on σ and τ .

Then as $h \rightarrow 0$, $\rho(D^{-1}C) = 1 - \Lambda_0 h^2 + o(h^2)$, where Λ_0 is the **smallest eigenvalue** of the problem

$$(5.2) \quad \mathbf{A}u = Aq u \quad \text{in } \Omega, \quad u = 0 \text{ on } \partial\Omega.$$

In assumption (PS4), \mathbf{e}_1 is a function of **first** order **differences** in u and u and \mathbf{e}_2 is a **function of second order differences**; see [14] for a more precise statement.

By Theorem 1, the reduced matrix is an appropriate approximation to A . Condition (PS1) has been **established** in §3 and §4. For both the one-line and **two-line** Jacobi splittings, condition (PS2) follows from the **Perron-Frobenius** theory, using the **fact** that D is an **M -matrix** for all small enough h [4]. Condition (PS3) follows from Lemma 3. Thus, it remains to determine q for condition (PS4). Much of [12] and [14] is concerned with how to do this. In particular, §7 of [12] and §9 of [14] imply that $q = 1$ for the one-line Jacobi splitting for the reduced system and $q = 3/4$ for the **two-line** Jacobi splitting. It is straightforward to **verify that** the **eigenfunctions** and **eigenvalues** of (5.2) for $q = 1$ are

$$u^{(j,k)} = e^{\sigma x/2} \sin(j\pi x) e^{\tau y/2} \sin(k\pi y), \quad \Lambda_{jk} = \frac{\sigma^2}{4} + \frac{\tau^2}{4} + (j^2 + k^2)\pi^2,$$

for integers $j, k \geq 1$. The minimum **eigenvalue** is $\Lambda_0 = \frac{\sigma^2}{4} + \frac{\tau^2}{4} + 2\pi^2$. Hence, we have the following asymptotic result (which applies **for** both difference schema):

COROLLARY 4. The **spectral** radii of the block **Jacobi** iteration **matrices** for the one-line **orderings** of the reduced **system** are bounded by

$$1 - \left(\frac{\sigma^2}{4} + \frac{\tau^2}{4} + 2\pi^2 \right) h^2 + o(h^2),$$

and the spectral radii of the block Jacobi iteration matrices for the two-line orderings are bounded by

$$1 - \left(\frac{\sigma^2}{3} + \frac{\tau^2}{3} + \frac{8}{3}\pi^2 \right) h^2 + o(h^2).$$

For large σ and τ (and small enough h), these bounds are essentially of the form $1 - c(\gamma^2 + \delta^2)$.

The analyses of §3 and §4 give asymptotic bounds of

$$1 - \left(\frac{\sigma^2}{4} + \frac{\tau^2}{4} + \frac{\pi^2}{4} \right) h^2, \quad 1 - \left(\frac{\sigma^2}{3} + \frac{\tau^2}{3} + 2\pi^2 \right) h^2,$$

for the one-line and **two-line** block Jacobi iteration matrices, respectively. These results agree with those of Corollary 4 except in the coefficient of π^2 . They are pessimistic because the numerators and denominators come from separate bounds, and (for the one-line case) because Gershgorin's theorem is used for the numerator. However, it may be more important to know the spectral radius in the nonasymptotic regime, i.e. for particular values of γ and δ not close to zero. The numerical experiments of §6 below indicate that the bounds of §3 and §4 are good indicators of spectral radii in such cases.

Note that smaller values of q in Theorem 7 produce smaller spectral radii. The analysis of [14] shows that for the 1-line Jacobi splitting of the unreduced system (which gives rise to methods comparable in cost to both methods considered here for the reduced system), $q = 2$. Thus, the asymptotic value of the spectral radius is smaller for the reduced system. An alternative proof of this fact, derived from regular splitting arguments (which are less dependent on **asymptotics**) is given in [6]. This observation is in agreement with results on spectral radii in [4]. Thus, asymptotic convergence behavior will be worse for the **full** system.

6. Numerical experiments and implementation.

In this section, we present the results of numerical experiments that **confirm** and supplement the analysis of §§3 – 4. For the two-line ordering, we compare the bounds on spectral radii of iteration matrices with computed spectral radii, and for all the orderings considered, we examine the performance of the Gauss-Seidel and SOR methods for solving the reduced **system** arising from the centered difference **discretization** of the **convection-diffusion equation**. Except where indicated, all computations were performed on a VAX-8600 in double precision **Fortran**. The reduced matrices were computed using PCGPAK [15]. All **spectral** radii were computed using the **QZ** algorithm in EISPACK [7], [8].

6.1. Spectral radii for the two-line methods.

Tables 6.1 – 6.3 show the computed **values** of the spectral radii of the Gauss-Seidel iteration matrices for the two-line orderings, for three values of h and **different** choices of the parameters γ and δ . In addition, the last column of each table shows the asymptotic limits (as $h \rightarrow 0$) of the bounds on these spectral radii, when such a bound **exists**. For $|\gamma|$,

7	$h = 1/8$	$h = 1/16$	$h = 1/32$	Asymptotic Bound
.2	.42	.74	.86	.90
.4	.33	.55	.63	.66
.6	.22	.34	.38	.40
.8	.11	.16	.18	.19
1.0	.01	.02	.06 ³	.02
1.2	.03	.04	.04	
1.4	.05	.06	.06	
1.6	.06	.06	.07	
1.8	.07	.07	.07	
2.0	.07	.07	.07	
3.0	.07	.07	.07	

Table 6.1: Spectral radii and bounds for the **two-line** Gauss-Seidel iteration matrices, centered differences, $\delta = 0$.

$|\delta| \leq 1$, these quantities are the squares of the limiting values from Corollary 2, where the **values for γ or $\delta = 1$** are the limits as $\gamma, \delta \rightarrow 1$. For Table 6.3 when $|\gamma| > 1$, we use Theorem 7. As in [4], the experimental results show that the bounds are good approximations to the limits as $h \rightarrow 0$ when $|\gamma| \leq 1$ and $|\delta| \leq 1$, and the bounds for $|\gamma|, |\delta| > 1$ are pessimistic. For values of 7 and δ where the analysis does not apply, the computed spectral radii are very close to zero. Note that the asymptotic results are expressed in a nonstandard way. In contrast to the **analysis** of §5, γ and δ are **fixed** here as $h \rightarrow 0$, so the continuous problem (1.4) is varying.

6.2. Performance of the block iterative methods.

Figs. 6.1 – 6.3 **summarize** the **performance** of the block iterative methods for solving various examples of the discrete convection-diffusion equation (1.4) with Dirichlet boundary conditions. In all cases, centered differences were used to discretize the **first** derivative terms, and the mesh size was $h = 1/32$, so that the order of the linear system was $N = 961$. The curves in the figures represent the average iteration counts for three test problems, determined by **three** initial guesses with random values in the interval $[-1, 1]$. In all cases, the right hand side **s was** identically **zero**. The convergence criterion was $\|r_i\|_2/\|r_0\|_2 \leq 10^{-6}$, where $r_i = s - Su_i^{(b)} = -Su_i^{(b)}$ is the residual at the i 'th iteration.

The left side of each of these **figures** contains results for the one-line orderings, and the right side contains results for the **two-line** orderings. Experiments were run for values of γ or δ equal to **multiples** of 0.2 in $[0, 2]$, plus 7 (or 6) = 3. Fig. 6.1 corresponds to the case $\delta = 0$ (i.e. only the u_x **first** order term was **present** in (1.4)), Fig. 6.2 to $7 = 0$

³ This computed **spectral radius exceeds** the analytic bound. **Computations** on a Sun 3/60 gave the same **results**. We believe that **this eigenvalue** computation **is** affected by ill-conditioning, although we do not understand the difficulty.

δ	$h = 1/8$	$h = 1/16$	$h = 1/32$	Asymptotic Bound
.2	.42	.74	.85	.90
.4	.32	.54	.62	.65
.6	.19	.30	.34	.36
.8	.07	.11	.12	.12
1.0	0	0	0	0
1.2	.03	.04	.04	
1.4	.06	.08	.09	
1.6	.09	.13	.13	
1.8	.12	.16	.18	
2.0	.14	.20	.22	
3.0	.21	.36	.41	

Table 8.2: Spectral radii and bounds for the **two-line** Gauss-Seidel iteration matrices, centered differences, $\gamma = 0$.

γ	$h = 1/8$	$h = 1/16$	$h = 1/32$	Asymptotic Bound
.2	.39	.67	.77	.81
.4	.23	.37	.42	.44
.6	.09	.14	.16	.16
.8	.02	.03	.03	.03
1.0	0	0	0	0
1.2	.01	.02	.02	.03
1.4	.04	.05	.05	.13
1.6	.08	.09	.09	.34
1.8	.12	.12	.12	.71
2.0	.16	.16	.16	1.27
3.0	.32	.33	.33	9.00

Table 6.3: Spectral radii and bounds for the **two-line** Gauss-Seidel iteration matrices, centered differences, $\gamma = 6$.

(only \mathbf{u}_y), and Fig. 6.3 to $\gamma = \delta$ (\mathbf{u}_x and \mathbf{u}_y). The results are for the block Gauss-Seidel method with **the** natural, red-black and torus orderings. (The iteration matrices for the alternating torus ordering are similar, via permutation matrices, to **those** for the one-line red-black ordering, so that these orderings produce identical iterates.) In addition, results for the block SOR method with the natural **ordering** are shown for some choices of γ and 6. For SOR, we used the optimal value of w determined by (3.3), where $\rho(\mathbf{B})^2$ is taken **from** Tables 6.1 – 6.3 and analogous results **from** [4], using the values for $h = 1/32$.

We make the following observations concerning these results:

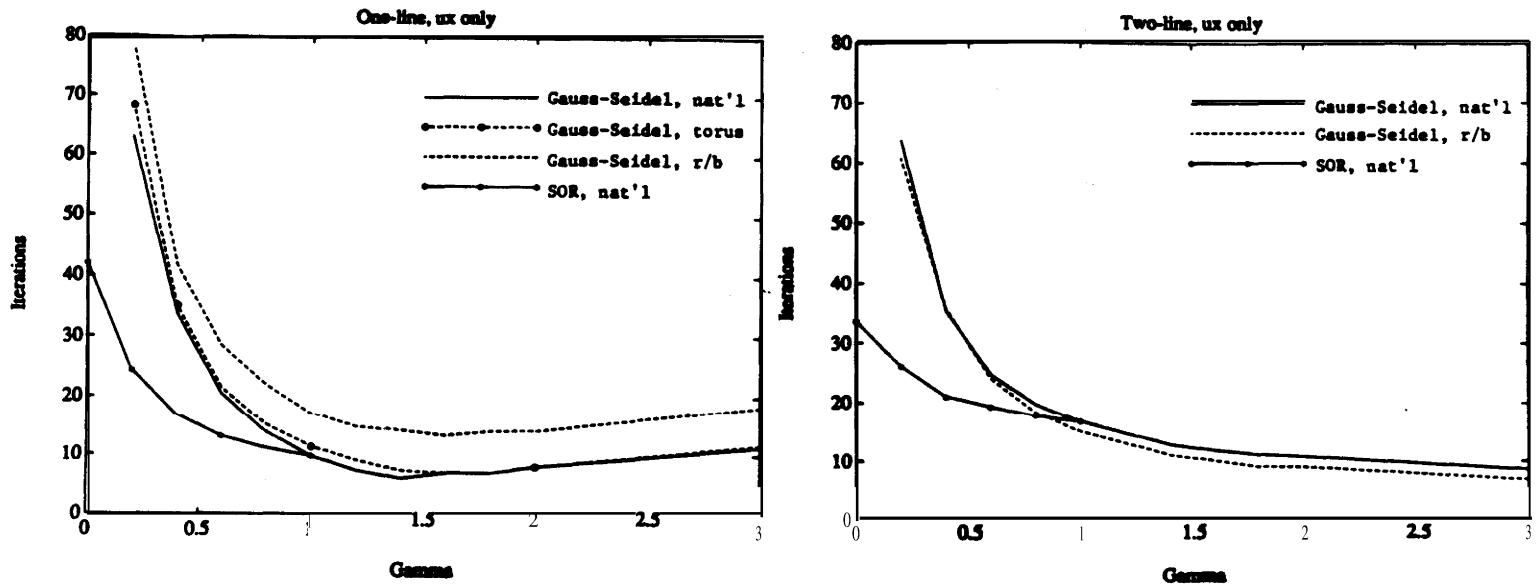


Fig. 6.1: Average iteration counts, $h = 1/32$, $6 = 0$.

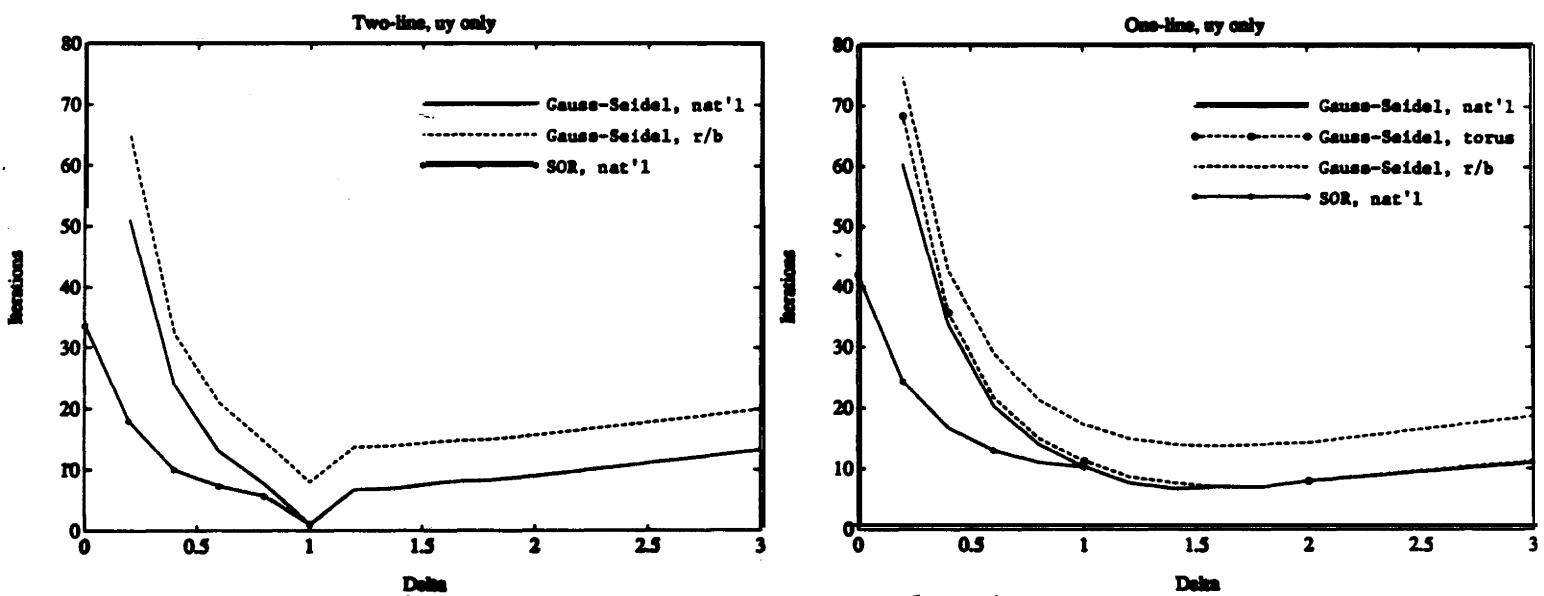


Fig. 8.3: Average iteration counts, $h = 1/32$, $\gamma = 0$.

- (1) In most cases, the Gauss-Seidel method requires thirty or fewer iterations to reach the stopping criterion. In general, fewer iterations are required with the natural **orderings than** with the red-black orderings; a rough estimate is that the **red-black orderings entail** at most twice as many iterations **as** the natural orderings. An exception **is** when $\delta = 0$, where the **performances** of the natural and red-black **two-line orderings are** very **close** (see the right side of Fig. 6.1).
- (2) The best results are obtained when γ or δ are near one, and performance typically improves as $|\gamma|$ or $|\delta| \rightarrow 1$. For all values of γ and δ tested, the self-adjoint case ($\gamma = \delta = 0$) required the largest number of Gauss-Seidel iterations. In these cases, for which 'the results are not shown on the figures, the stopping criterion was typically not reached after 150 iterations. In general, performance is in

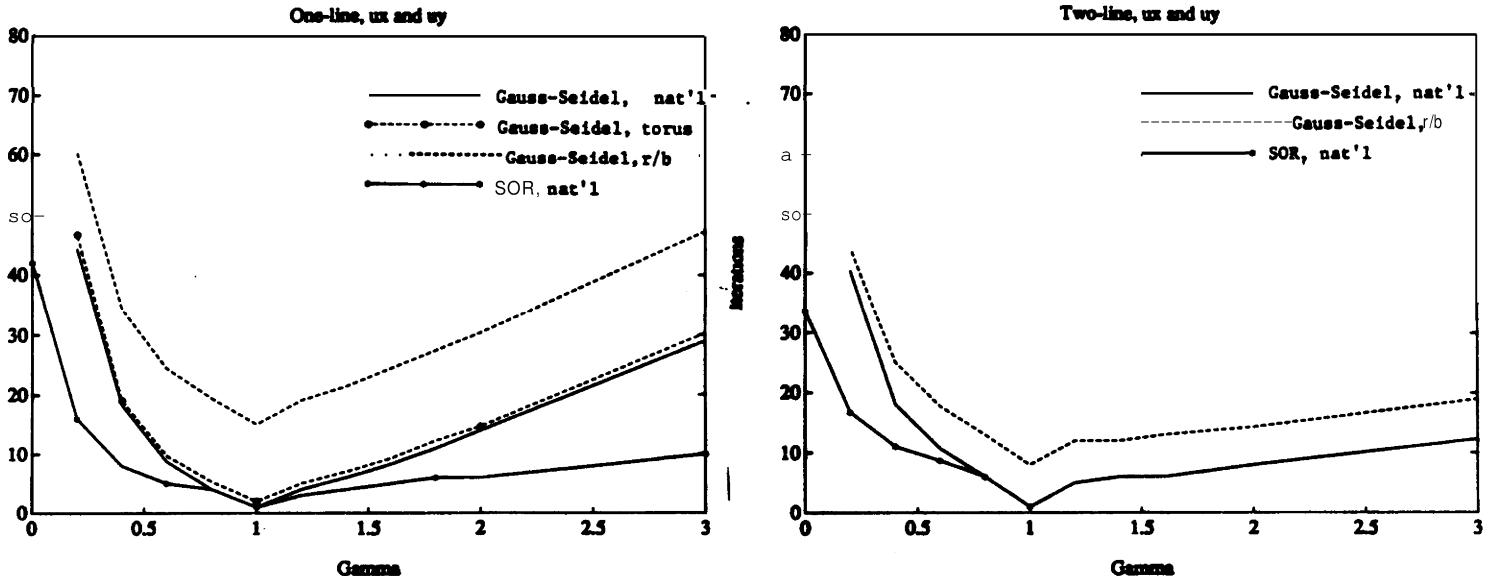


Fig. 6.3: Average iteration counts, $h = 1/32$, $7 = 6$.

accordance with the results on spectral radii from Tables 6.1 – 6.3 and [4].

- (3) The best results for large 7 or δ are for the **two-line** orderings with $\delta = 0$ (Table 6.1 and Fig. 6.1). This is because as $|\gamma|$ grows, S essentially consists of its block diagonal \mathbf{D} plus a small perturbation. For large δ and $7 = 0$, a **vertical two-line** splitting would give better results than the horizontal splitting used.
- (4) SOR was much more **effective** than Gauss-Seidel when the latter was slow. We examined SOR only in cases where the spectrum of the block Jacobi iteration matrix is real, i.e. where either $|\gamma| < 1$ and $|\delta| < 1$ or (for the one-line ordering [4]) $|\gamma| > 1$ and $|\delta| > 1$. Thus, (3.3) applies. In variable **coefficient** problems of a similar character, it would be realistic to use an adaptive method to estimate the optimal value of w (see e.g. [18]). For other values of 7 or 6 , the spectral radius of the Gauss-Seidel iteration matrix is already very small, and we did not experiment with **SOR**. To keep the graphs from being too detailed, the SOR results are shown only for the natural orderings. Like Gauss-Seidel, with the **red-black** orderings SOR typically required somewhat more iterations, but it displayed the same general character **as** it did with the natural ordering (i.e. graphs of iteration counts have similar slopes).
- (5) The performance of the Gauss-Seidel method with the torus ordering is very close to its performance with the natural one-line ordering.

The error $\mathbf{e}_j \equiv \mathbf{u} - \mathbf{u}_j$ at the j 'th step of each of the methods under consideration satisfies $\mathbf{e}_j = \mathbf{M}\mathbf{e}_{j-1}$, where \mathbf{M} is the iteration matrix. Thus, for large enough j , the error will be dominated by the eigenvector corresponding to the spectral radius, and the asymptotic (in terms of iteration counts) analysis of §3 and §4 can be used to predict behavior. However, this does not say anything about **how other** components of the error **affect** performance, and it also does not explain the effects of different orderings. Figures

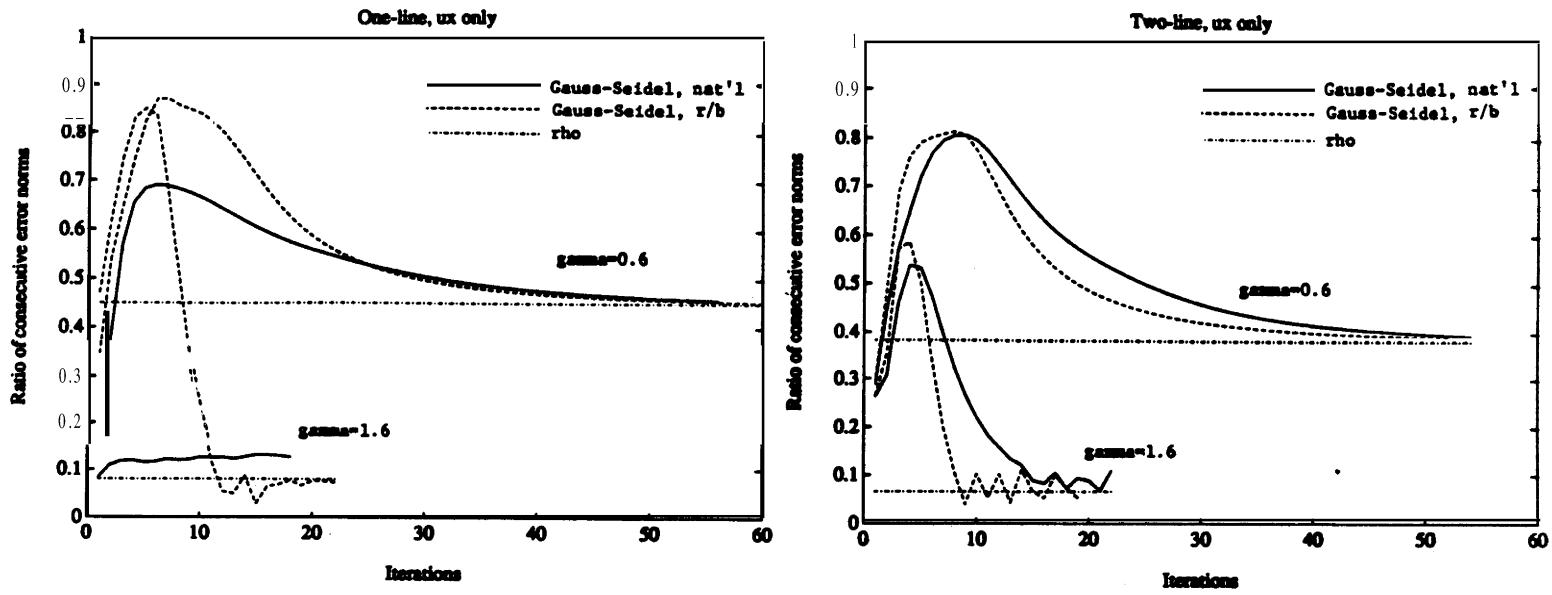


Fig. 6.4: Approach to asymptotic performance, $h = 1/32, \delta = 0$.

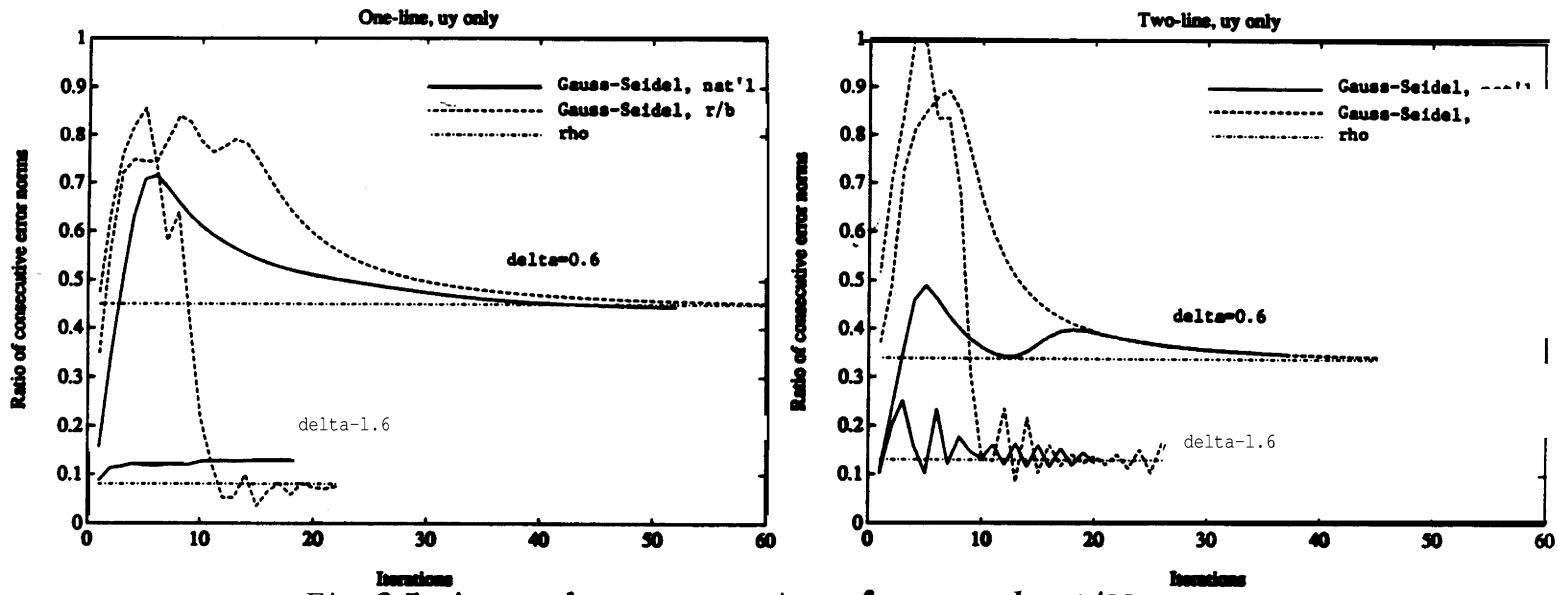


Fig. 6.5: Approach to asymptotic performance, $h = 1/32, \gamma = 0$.

6.4 – 6.6 examine the question of when the asymptotic behavior takes **effect**, in the **Gauss-Seidel** method. Each figure graphs the ratio $\|e_j\|_2/\|e_{j-1}\|_2$, for the natural and red-black versions of both the one-line and two-line orderings, for two problems, one where γ or δ is **less** than one, and one where γ or δ is greater than one. Figure 6.4 shows the **case** where $\gamma = .6$ and 1.6 **and** $\delta = 0$; Figure 6.5 shows the case where $\gamma = 0$, and $\delta = .6$ and 1.6 ; and Figure 6.6 shows the case where $\gamma = \delta = .6$ and 1.6 . These results are for one of the initial guesses used in the experiments described above. In **all** cases, the iterations were performed until the (stringent) stopping criterion $\|e_i\|_2/\|e_0\|_2 \leq 10^{-16}$ was satisfied.

The **results** show that the behavior of the Gauss-Seidel method is typically closer to that predicted by the asymptotic analysis when the natural ordering is used, and that fewer iterations are required before the asymptotic performance is seen. The one exception in these examples is Where $\delta = 0$ with the two-line ordering (Fig. 6.4); in this case the

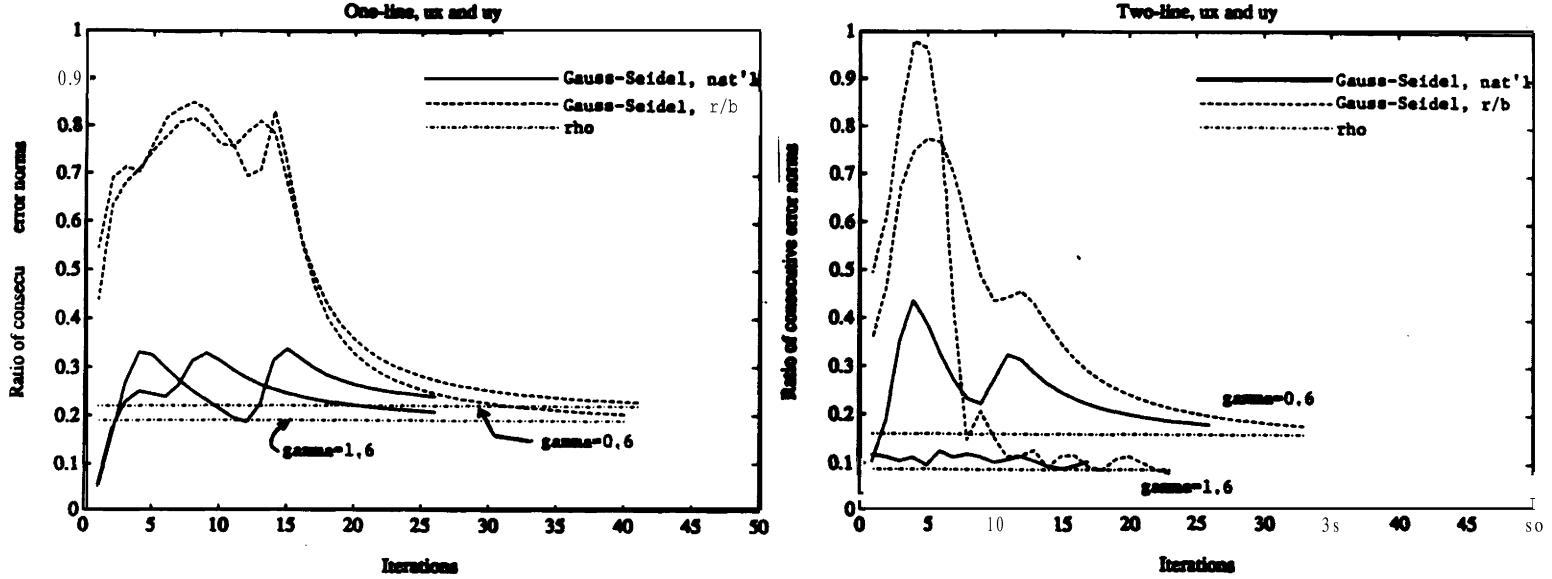


Fig. 6.6: Approach to asymptotic performance, $h = 1/32$, $\gamma = \delta$.

natural and red-black orderings display similar asymptotic behavior. Recall that this was the one case where **the performances** were similar. We also remark that the asymptotic performance is typically displayed only after the stopping criterion used for **Figs. 6.1 – 6.3** is satisfied.

		One-line Natural R/B		Two-line Natural R/B	
$\delta = 0$	$\gamma = 0.6$.86	1.38	1.12	1.35
	$\gamma = 1.6$.27	1.40	1.00	1.27
$\gamma = 0$	$\delta = 0.6$.86	1.38	.92	1.47
	$\delta = 1.6$.27	1.40	1.57	1.65
$\gamma = \delta$	$\gamma = 0.6$.53	1.40	.87	1.46
	$\gamma = 1.6$.53	1.40	1.14	1.65

Table 6.4: Euclidean norms of the Gauss-Seidel iteration matrices.

I With $M = \mathcal{L}_1$, the errors for the Gauss-Seidel iteration satisfy $e_j = \mathcal{L}_1^j e_0$, so that $\|\mathcal{L}_1^j\|$ would give more precise predictions of the behavior of the errors. Table 6.4 shows $\|\mathcal{L}_1\|_2$ for the twelve examples of Figs. 6.4 – 6.6. These norms were computed by taking the maximum singular **value**, acquired using LINPACK [3] (in double precision **Fortran**) on a SUN 3/60. The results show that the norms for the natural orderings are typically less than one, and the norms for the red-black ordering are typically greater than one **as well as** greater than those for the natural ordering. Thus, the results are largely consistent with the numerical behavior described above. There are cases, however, where $\|\mathcal{L}_1\|_2 > 1$ but the asymptotic behavior is good, e.g. $\gamma = 0$, $\delta = 1.6$, two-line natural ordering (see Fig 6.5).

6.3. Implementation and parallelism.

We now outline the implementation costs of the block iterative methods for solving the reduced system. We focus on the block SOR iteration, of which the Gauss-Seidel method is a special case. Assume that the **reduced matrix has** the form S_{ij} , where the indices refer to the blocks associated with the lines of the ordering in use. For example, for the natural **two-line** ordering, i and j vary between 1 and $n/2$ and $S_{ij} = 0$ for $|i - j| > 1$. Let $S = D - (L + U)$ where D , L and U are blocked in an analogous manner, and let s and $z \equiv u^{(b)}$ be **indexed** in an analogous manner. Note that each block of D is a banded matrix of total bandwidth either three (for the one-line orderings) or **five** (for the two-line orderings). Assume for simplicity that the LU-factorization of each D_i can be computed without pivoting. (This is the case whenever D diagonally dominant.)

The block SOR iteration has the form

$$(6.1) \quad z_i^{(m+1)} = z_i^{(m)} - \omega [z_i^{(m)} - D_i^{-1} \left(\sum_{j < i} L_{ij} z_j^{(m+1)} + \sum_{j > i} U_{ij} z_j^{(m)} \right)] + D_i^{-1} s_i,$$

where i varies from 1 to the number of blocks in the matrix. Consider the computations involving the matrices D , L and U . Each step requires a matrix-vector product by the i 'th block row of U and a matrix-vector product by the i 'th block row of L , followed by a linear solve in which the **coefficient** matrix is the i 'th block of D . The cost of the matrix-vector products (in terms of multiply-adds) is essentially equal to the number of nonzeros in the i 'th block rows of L and U . Moreover, assuming that D_i has been factored, the cost of the linear solve is equal to the number of nonzeros in D_i . Consequently, for any of the orderings, the total cost of the matrix computations on a serial computer is approximately **$9n^2/2$** , the number of **nonzeros** in S . All the other computations (vector adds and **scalar**-matrix products) are clearly independent of ordering. The factorization of the blocks of D is slightly more expensive for the two-line ordering than for the one-line ordering, but both are of the order of the cost of one iteration, so that the difference is negligible. Pivoting will have a somewhat more detrimental **effect** on the two-line orderings than on the one-line orderings.

Both the natural and red-black orderings have **efficient** implementations on parallel computers with $k = O(n)$ processors. The architecture need not have a more complex topology than a linear array (or a ring for the torus orderings), and our discussion applies as **well** to shared memory machines. It is straightforward to show that the construction of the reduced **system** is fully **parallelizable**. In examining the iterative methods, we assume for simplicity **that** the ordering is such that all block rows of the reduced matrix are of the same size. **This** is the case for the torus **one-line** ordering and for the two-line ordering when n is even; the size is **approximately** n . Let n_r denote the number of block rows; for all orderings, $n_r \approx n/2$. Assume further that k divides n_r , and let the processors be indexed from 1 to k .

The iterations for the natural **versions** of these orderings can be pipelined using the methods of [1] where a (block) **step** of the computation is **defined** by the following rule:

at the i 'th step, Processor j is performing the $(i - j + 1)$ 'st iteration on the **first** $j \times n_r/k$ block rows.

That is, at step one, Processor 1 performs the **first** iteration on the **first n_r/k** block rows. Then, at step 2, Processor **2** performs the **first** iteration on the second n_r/k block rows, and Processor 1 performs the second iteration on the **first n_r/k** block rows. The **first** iteration is completed by Processor k after k such steps, and every subsequent step results in the completion of one more iteration. All processors are busy except during the **first** and **last $k - 1$** steps. For t iterations, the **speedup** (&arithmetic) is

$$\frac{k}{1 + k/t}.$$

Thus, the pipelined implementation is efficient whenever t is large relative to k . For architectures with distributed memory, neighboring processors must exchange vectors of length (approximately) n *between step*, and some overlap of communication and arithmetic is possible.

The alternating torus ordering requires that $\lceil n/2 \rceil$ be even in order to correspond to a red-black ordering; no additional assumptions on n are needed for the two-line red-black ordering. Both red-black orderings are then fully parallelizable on up to $n/4$ processors. For all indices i with red color, the computation (6.1) consists of a set of independent block matrix-vector products by the **nonzero** blocks of \mathbf{U} , followed by a set of independent block matrix solves. Then, for all indices i with black color, the steps of (6.1) consist of a set of independent block matrix-vector products by the **nonzero** blocks of L , followed by a set of independent block matrix solves. Unidirectional communication between neighboring processors of vectors of length n is needed twice, prior to the multiplications by L and \mathbf{U} . Overlap with arithmetic is possible.

In §6.2, we found the methods to be very **effective** on model problems. For the **small values** of t **observed**, it appears that the **inefficiency** of the natural orderings due to pipelining will often be similar in scale to the somewhat slower performance displayed by the red-black orderings. Consequently, we expect the performance of the two classes of orderings to be comparable on parallel architectures.

7. Concluding remarks.

In this paper, we have continued the analysis begun in [4] of block iterative methods for solving cyclically reduced linear systems derived **from** the convection-diffusion equation. We showed how the discrete reduced system is related to the underlying continuous problem, and we derived bounds on the spectral radius of the block Jacobi iteration matrix **associated with** two-line orderings of the reduced grid. These bounds, together with analogous **ones from** [4], were **combined** with the Young theory to analyze the asymptotic convergence behavior of the Gauss-Seidel and SOR block iterative methods derived **from** several **variants** of both two-line orderings and one-line orderings. The results express convergence behavior in terms of discrete cell Reynolds numbers $\sigma h/2, \tau h/2$, and they are **confirmed** and supplemented by numerical experiments. The analytic and experimental results (as well as those of [2] and [14]) show that the nonsymmetric discrete problems arising from (1.4) are in some ways easier to solve than the symmetric ones.

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