

On solving linear systems arising from Shishkin mesh discretizations

Petr Tichý

Faculty of Mathematics and Physics, Charles University

joint work with

Carlos Echeverría, Jörg Liesen, and Daniel Szyld

October 20, 2016, Prague

Seminar of Numerical Mathematics, KNM MFF UK

Problem formulation

Convection-diffusion boundary value problem

$$\begin{aligned} -\epsilon u'' + \alpha u' + \beta u &= f \quad \text{in } \Omega = (0, 1), \\ u(0) &= u_0, \quad u(1) = u_1, \end{aligned}$$

$\alpha > 0$, $\beta \geq 0$ constants, the problem is **convection dominated**

$$0 < \epsilon \ll \alpha.$$

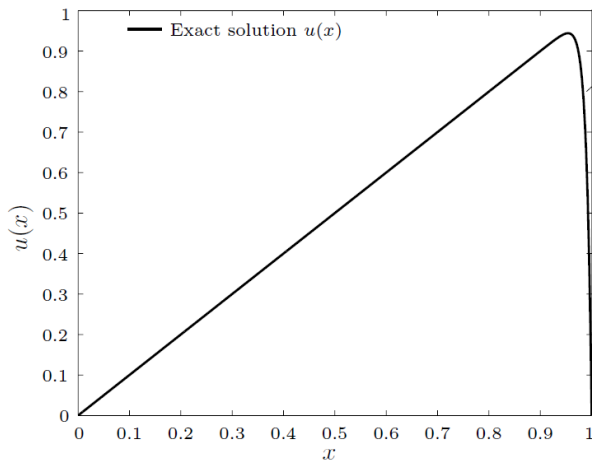
[Stynes, 2005] (Acta Numerica)

[Roos, Stynes, and Tobiska, 1996, 2008] (book)

[Miller, O'Riordan, and Shishkin, 1996] (book)

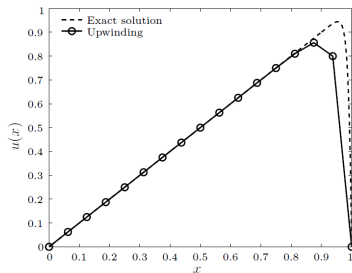
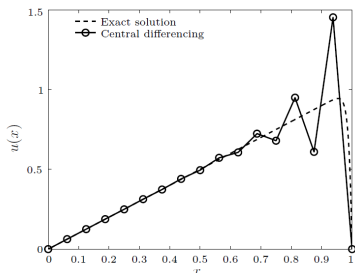
Solution and boundary layers

$$\epsilon = 0.01, \alpha = 1, \beta = 0, u(0) = u(1) = 0.$$



There are small subregions where the solution has a large gradient.

Numerical solution, equidistant mesh



Standard techniques:

$$u'(ih) \approx \frac{u_{i+1} - u_{i-1}}{2h}, \quad u'(ih) \approx \frac{u_i - u_{i-1}}{h}.$$

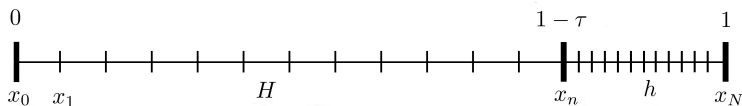
- Unnatural **oscillations** or **cannot resolve** the layers.
- Remedy: stabilization or **non-equidistant** mesh.
- We study discretizations for a **Shishkin mesh**.

Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?
- 3 Multiplicative Schwarz method
- 4 Convergence analysis
- 5 Schwarz method as a preconditioner
- 6 Numerical examples

Shishkin mesh on $[0, 1]$

Piecewise equidistant



N even, define the **transition point** $1 - \tau$ and n by

$$\tau \equiv \min \left\{ \frac{1}{2}, \frac{\epsilon}{\alpha} 2 \ln N \right\}, \quad n \equiv \frac{N}{2}.$$

If $\epsilon \ll \alpha$, then $1 - \tau$ is close to 1. Next define H and h by

$$H \equiv \frac{1 - \tau}{n}, \quad h \equiv \frac{\tau}{n}$$

and consider the Shishkin mesh, $x_0 = 0$,

$$x_i \equiv iH, \quad x_{n+i} \equiv x_n + ih, \quad i = 1, \dots, n.$$

Discretization on the Shishkin mesh - details

For simplicity $u(0) = u(1) = 0$

The **upwind** difference scheme is given by

$$-\epsilon \delta_x^2 u_i + \alpha D_x^- u_i + \beta u_i = f_i, \quad u_0 = u_N = 0,$$

and the **central difference** scheme by

$$-\epsilon \delta_x^2 u_i + \alpha D_x^0 u_i + \beta u_i = f_i, \quad u_0 = u_N = 0,$$

where

$$\delta_x^2 u_i = \frac{2u_{i-1}}{(H+h)H} - \frac{2u_i}{Hh} + \frac{2u_{i+1}}{(H+h)h}, \quad i = n,$$

and

$$D_x^- u_i = \frac{u_i - u_{i-1}}{H}, \quad 1 \leq i \leq n, \quad D_x^0 u_i = \frac{u_{i+1} - u_{i-1}}{h + H}, \quad i = n.$$

surveys [Linss, Stynes, 2001], [Stynes, 2005], [Kopteva, O'Riordan, 2010]

Shishkin mesh and ϵ -uniform convergence

- The **upwind** difference scheme

$$-\epsilon \delta_x^2 u_i + \alpha D_x^- u_i + \beta u_i = f_i, \quad u_0 = u_N = 0.$$

There exists a constant C such that

$$|u(x_i) - u_i| \leq C \left(\frac{\ln N}{N} \right), \quad i = 0, \dots, N,$$

see, e.g., [Stynes, 2005].

Shishkin mesh and ϵ -uniform convergence

- The **upwind** difference scheme

$$-\epsilon \delta_x^2 u_i + \alpha D_x^- u_i + \beta u_i = f_i, \quad u_0 = u_N = 0.$$

There exists a constant C such that

$$|u(x_i) - u_i| \leq C \left(\frac{\ln N}{N} \right), \quad i = 0, \dots, N,$$

see, e.g., [Stynes, 2005].

- The **central** difference scheme

$$-\epsilon \delta_x^2 u_i + \alpha D_x^0 u_i + \beta u_i = f_i, \quad u_0 = u_N = 0.$$

There exists a constant C such that

$$|u(x_i) - u_i| \leq C \left(\frac{\ln N}{N} \right)^2, \quad i = 0, \dots, N,$$

[Andreev and Kopteva, 1996], a difficult proof, the scheme does not satisfy a discrete maximum principle. [Kopteva and Linss, 2001].

Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?**
- 3 Multiplicative Schwarz method
- 4 Convergence analysis
- 5 Schwarz method as a preconditioner
- 6 Numerical examples

Structure of the matrix

$$A = \left[\begin{array}{ccc|c|c} a_H & b_H & & & \\ c_H & \ddots & \ddots & & \\ & \ddots & \ddots & b_H & \\ & & c_H & a_H & b_H \\ \hline & & & c & a & b \\ \hline & & & c_h & a_h & b_h \\ & & & & \ddots & \ddots \\ & & & & \ddots & \ddots & b_h \\ & & & & & c_h & a_h \end{array} \right]$$

The upwind scheme

$$\begin{aligned}
 c_H &= -\frac{\epsilon}{H^2} - \frac{\alpha}{H}, & a_H &= \frac{2\epsilon}{H^2} + \frac{\alpha}{H} + \beta, & b_H &= -\frac{\epsilon}{H^2}, \\
 c &= -\frac{2\epsilon}{H(H+h)} - \frac{\alpha}{H}, & a &= \frac{2\epsilon}{hH} + \frac{\alpha}{H} + \beta, & b &= -\frac{2\epsilon}{h(H+h)}, \\
 c_h &= -\frac{\epsilon}{h^2} - \frac{\alpha}{h}, & a_h &= \frac{2\epsilon}{h^2} + \frac{\alpha}{h} + \beta, & b_h &= -\frac{\epsilon}{h^2}.
 \end{aligned}$$

The central difference scheme

$$\begin{aligned}
 c_H &= -\frac{\epsilon}{H^2} - \frac{\alpha}{2H}, & a_H &= \frac{2\epsilon}{H^2} + \beta, & b_H &= -\frac{\epsilon}{H^2} + \frac{\alpha}{2H}, \\
 c &= -\frac{2\epsilon}{H(H+h)} - \frac{\alpha}{h+H}, & a &= \frac{2\epsilon}{hH} + \beta, & b &= -\frac{2\epsilon}{h(H+h)} + \frac{\alpha}{h+H}, \\
 c_h &= -\frac{\epsilon}{h^2} - \frac{\alpha}{2h}, & a_h &= \frac{2\epsilon}{h^2} + \beta, & b_h &= -\frac{\epsilon}{h^2} + \frac{\alpha}{2h}.
 \end{aligned}$$

Matrix properties

- nonsymmetric
- A is M-matrix for the upwind scheme
- A is not an M-matrix for the central difference scheme
- A is highly **nonnormal**. Consider

$$-\epsilon u'' + u' = 1, \quad u(0) = 0, \quad u(1) = 0,$$

$\epsilon = 10^{-8}$ and $N = 46$, and the spectral decomposition

$$A = YDY^{-1}.$$

Matrix properties

- nonsymmetric
- A is M-matrix for the upwind scheme
- A is not an M-matrix for the central difference scheme
- A is highly **nonnormal**. Consider

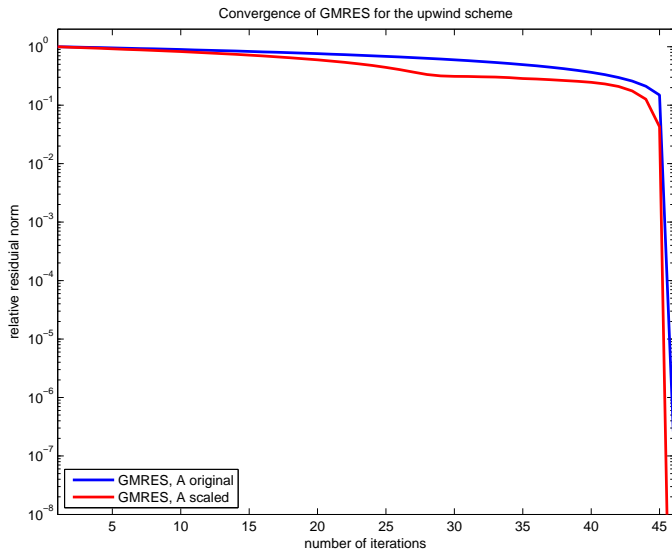
$$-\epsilon u'' + u' = 1, \quad u(0) = 0, \quad u(1) = 0,$$

$\epsilon = 10^{-8}$ and $N = 46$, and the spectral decomposition

$$A = YDY^{-1}.$$

| | upwind | upwind sc. | central | central sc. |
|-------------|-----------------------|-----------------------|-----------------------|--------------------|
| $\kappa(A)$ | 4.05×10^{10} | 2.96×10^3 | 6.23×10^{10} | 2.95×10^3 |
| $\kappa(Y)$ | 1.51×10^{17} | 1.23×10^{19} | 4.11×10^3 | 1.87×10^2 |

Solving linear system using GMRES



Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?
- 3 Multiplicative Schwarz method**
- 4 Convergence analysis
- 5 Schwarz method as a preconditioner
- 6 Numerical examples

Multiplicative Schwarz method

Idea of solving $Ax = b$

- Given an approximation $x^{(k)}$, then $x = x^{(k)} + y$ and y satisfies

$$Ay = b - Ax^{(k)}.$$

Multiplicative Schwarz method

Idea of solving $Ax = b$

- Given an approximation $x^{(k)}$, then $x = x^{(k)} + y$ and y satisfies

$$Ay = b - Ax^{(k)}.$$

- Restriction** operators $R_1 = \begin{bmatrix} I_n & 0 \end{bmatrix}$, $R_2 = \begin{bmatrix} 0 & I_n \end{bmatrix}$.

Multiplicative Schwarz method

Idea of solving $Ax = b$

- Given an approximation $x^{(k)}$, then $x = x^{(k)} + y$ and y satisfies

$$Ay = b - Ax^{(k)}.$$

- Restriction** operators $R_1 = \begin{bmatrix} I_n & 0 \end{bmatrix}$, $R_2 = \begin{bmatrix} 0 & I_n \end{bmatrix}$.
- Solve on the **first domain**

$$(R_1 A R_1^T) \tilde{y} = R_1 (b - Ax^{(k)})$$

and approximate y by **prolongation** of \tilde{y} , i.e., by $R_1^T \tilde{y}$. Define

$$x^{(k+\frac{1}{2})} = x^{(k)} + R_1^T (R_1 A R_1^T)^{-1} R_1 (b - Ax^{(k)}).$$

Multiplicative Schwarz method

Idea of solving $Ax = b$

- Given an approximation $x^{(k)}$, then $x = x^{(k)} + y$ and y satisfies

$$Ay = b - Ax^{(k)}.$$

- Restriction** operators $R_1 = \begin{bmatrix} I_n & 0 \end{bmatrix}$, $R_2 = \begin{bmatrix} 0 & I_n \end{bmatrix}$.
- Solve on the **first domain**

$$(R_1 A R_1^T) \tilde{y} = R_1 (b - Ax^{(k)})$$

and approximate y by **prolongation** of \tilde{y} , i.e., by $R_1^T \tilde{y}$. Define

$$x^{(k+\frac{1}{2})} = x^{(k)} + R_1^T (R_1 A R_1^T)^{-1} R_1 (b - Ax^{(k)}).$$

- Similarly, use $x^{(k+\frac{1}{2})}$ on the **second domain** and prolong,

$$x^{(k+1)} = x^{(k+\frac{1}{2})} + R_2^T (R_2 A R_2^T)^{-1} R_2 (b - Ax^{(k+\frac{1}{2})}).$$

Multiplicative Schwarz method

Formalism

Define

$$P_i = R_i^T A_i^{-1} R_i A, \quad A_i \equiv R_i A R_i^T, \quad i = 1, 2.$$

The multiplicative Schwarz is the iterative scheme

$$x^{(k+1)} = T x^{(k)} + v, \quad T = (I - P_2)(I - P_1),$$

where v is defined such that the scheme is **consistent**.

Multiplicative Schwarz method

Formalism

Define

$$P_i = R_i^T A_i^{-1} R_i A, \quad A_i \equiv R_i A R_i^T, \quad i = 1, 2.$$

The multiplicative Schwarz is the iterative scheme

$$x^{(k+1)} = T x^{(k)} + v, \quad T = (I - P_2)(I - P_1),$$

where v is defined such that the scheme is **consistent**. Hence,

$$x - x^{(k+1)} = T^{k+1} (x - x_0),$$

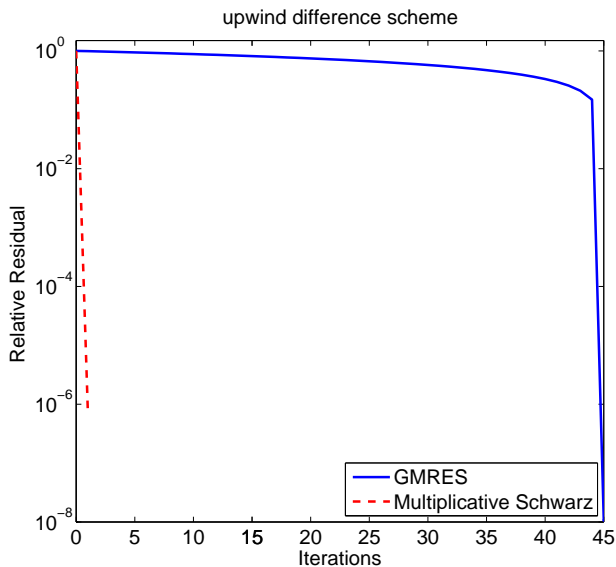
and

$$\|x - x^{(k+1)}\| \leq \|T^{k+1}\| \|x - x_0\|.$$

Is it convergent in our case?

Multiplicative Schwarz method

Experiment



Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?
- 3 Multiplicative Schwarz method
- 4 Convergence analysis**
- 5 Schwarz method as a preconditioner
- 6 Numerical examples

Convergence analysis

Exploiting the structure

$$\frac{\|x - x^{(k+1)}\|}{\|x - x_0\|} \leq \|T^{k+1}\|.$$

Using $T = (I - P_2)(I - P_1)$ we are able to show that

$$T = \left[\begin{array}{c|c|c} & t_1 & \\ & \vdots & \\ 0 \dots 0 & t_{n+1} & 0 \dots 0 \\ & \vdots & \\ & t_{N-1} & \end{array} \right] = t e_{n+1}^T.$$

Convergence analysis

Exploiting the structure

$$\frac{\|x - x^{(k+1)}\|}{\|x - x_0\|} \leq \|T^{k+1}\|.$$

Using $T = (I - P_2)(I - P_1)$ we are able to show that

$$T = \left[\begin{array}{c|c|c} & t_1 & \\ & \vdots & \\ 0 \dots 0 & t_{n+1} & 0 \dots 0 \\ & \vdots & \\ & t_{N-1} & \end{array} \right] = t e_{n+1}^T.$$

Therefore, $T^2 = t (e_{n+1}^T t) e_{n+1}^T = t_{n+1} T$, and

$$\|T^{k+1}\| = |t_{n+1}|^k \|T\|.$$

How to bound $|t_{n+1}|$, and $\|T\|$ in a convenient norm ($\|\cdot\|_\infty$)?

Convergence analysis

Details

$$A = \left[\begin{array}{c|c|c} A_H & & \\ \hline & b_H & \\ \hline c & a & b \\ \hline & c_h & A_h \end{array} \right].$$

Let $m \equiv n - 1$, $\rho \equiv |t_{n+1}| \dots$ the **convergence factor**. Then,

$$\rho = \left| \frac{bb_H (A_H^{-1})_{m,m}}{a - cb_H (A_H^{-1})_{m,m}} \right| \left| \frac{cc_h (A_h^{-1})_{1,1}}{a - bc_h (A_h^{-1})_{1,1}} \right|.$$

Convergence analysis

Bounding $(A_H^{-1})_{m,m}$ and $(A_h^{-1})_{1,1}$

A matrix $B = [b_{i,j}]$ is called a nonsingular **M -matrix** when

- B is nonsingular,
- $b_{i,i} > 0$ for all i , $b_{i,j} \leq 0$ for all $i \neq j$,
- and $B^{-1} \geq 0$ (elementwise).

Convergence analysis

Bounding $(A_H^{-1})_{m,m}$ and $(A_h^{-1})_{1,1}$

A matrix $B = [b_{i,j}]$ is called a nonsingular **M -matrix** when

- B is nonsingular,
- $b_{i,i} > 0$ for all i , $b_{i,j} \leq 0$ for all $i \neq j$,
- and $B^{-1} \geq 0$ (elementwise).

If A_H and A_h are nonsingular **M -matrices**, then using [Nabben 1999],

$$(A_H^{-1})_{m,m} \leq \min \left\{ \frac{1}{|b_H|}, \frac{1}{|c_H|} \right\},$$

$$(A_h^{-1})_{1,1} \leq \min \left\{ \frac{1}{|b_h|}, \frac{1}{|c_h|} \right\}.$$

A sufficient condition: The sign conditions & irreducibly diagonal dominant \Rightarrow nonsingular M -matrix. [Meurant, 1996], [Hackbusch, 2010]

Convergence analysis

The upwind scheme

The matrices A_H and A_h are M -matrices, and we know that

$$\frac{\|e^{(k+1)}\|_\infty}{\|e^{(0)}\|_\infty} \leq \rho^k \|T\|_\infty.$$

Convergence analysis

The upwind scheme

The matrices A_H and A_h are **M -matrices**, and we know that

$$\frac{\|e^{(k+1)}\|_\infty}{\|e^{(0)}\|_\infty} \leq \rho^k \|T\|_\infty.$$

Theorem (the upwind scheme)

[Echeverría, Liesen, T. , Szyld, 2016]

For the upwind scheme we have

$$\rho \leq \frac{\epsilon}{\epsilon + \alpha H} \leq \frac{\epsilon}{\epsilon + \frac{\alpha}{N}},$$

and

$$\|T\|_\infty \leq \frac{\epsilon}{\epsilon + \alpha H}.$$

Convergence analysis

The central difference scheme

- A_h is still an M -matrix.
- If $\alpha H > 2\epsilon$, i.e. $b_H > 0$, then A_H is not an M -matrix
... the most common situation from a practical point of view.

Convergence analysis

The central difference scheme

- A_h is still an M -matrix.
- If $\alpha H > 2\epsilon$, i.e. $b_H > 0$, then A_H is not an M -matrix
... the most common situation from a practical point of view.
- Recall

$$\rho = \left| \frac{bb_H(A_H^{-1})_{m,m}}{a - cb_H(A_H^{-1})_{m,m}} \right| \left| \frac{cc_h(A_h^{-1})_{1,1}}{a - bc_h(A_h^{-1})_{1,1}} \right|.$$

- How to bound $(A_H^{-1})_{m,m}$? ... results by [Usmani 1994]

Convergence analysis

The central difference scheme

- A_h is still an M -matrix.
- If $\alpha H > 2\epsilon$, i.e. $b_H > 0$, then A_H is not an M -matrix
... the most common situation from a practical point of view.
- Recall

$$\rho = \left| \frac{bb_H(A_H^{-1})_{m,m}}{a - cb_H(A_H^{-1})_{m,m}} \right| \left| \frac{cc_h(A_h^{-1})_{1,1}}{a - bc_h(A_h^{-1})_{1,1}} \right|.$$

- How to bound $(A_H^{-1})_{m,m}$? ... results by [Usmani 1994]
- We proved: If $m = N/2 - 1$ is even, then

$$b_H(A_H^{-1})_{m,m} \leq \frac{1 - \left| \frac{b_H}{c_H} \right|^m}{\left| \frac{c_H}{b_H} \right| + \left| \frac{b_H}{c_H} \right|^m} < \frac{2m\epsilon}{\epsilon + \frac{\alpha H}{2}}.$$

Convergence analysis

The central difference scheme

A_h is **M -matrix**, if $\alpha H > 2\epsilon$, A_H is not an M -matrix.

Theorem (the central diff. scheme) [Echeverría, Liesen, T. , Szyld, 2016]

Let $m = N/2 - 1$ be even, and let $\alpha H > 2\epsilon$. For the central differences we have

$$\rho < \frac{2m\epsilon}{\epsilon + \frac{\alpha H}{2}} < N \frac{\epsilon}{\epsilon + \frac{\alpha}{N}},$$

$$\|T\|_\infty < 2.$$

Thus, the error of the multiplicative Schwarz method satisfies

$$\frac{\|e^{(k+1)}\|_\infty}{\|e^{(0)}\|_\infty} < 2 \left(\frac{2m\epsilon}{\epsilon + \frac{\alpha H}{2}} \right)^k.$$

Remarks on diagonally scaled linear algebraic systems

$$DAx = Db$$

The ill-conditioning can be avoided by diagonal scaling [Roos 1996]:
 $Ax = b$ is multiplied from the left by

$$D = \left[\begin{array}{c|c|c} d_H I_m & & \\ \hline & d & \\ \hline & & d_h I_m \end{array} \right].$$

Remarks on diagonally scaled linear algebraic systems

$$DAx = Db$$

The ill-conditioning can be avoided by diagonal scaling [Roos 1996]:
 $Ax = b$ is multiplied from the left by

$$D = \left[\begin{array}{c|c|c} d_H I_m & & \\ \hline & d & \\ \hline & & d_h I_m \end{array} \right].$$

- Such a scaling **preserves** the **Toeplitz** structure and the **M -matrix** property of the submatrices.
- Analysis depends on these properties and on ratios between elements in the same row such as $|b/a|$ and $|b_H/c_H|$. These **ratios are invariant** under diagonal scaling.

Remarks on diagonally scaled linear algebraic systems

$$DAx = Db$$

The ill-conditioning can be avoided by diagonal scaling [Roos 1996]:
 $Ax = b$ is multiplied from the left by

$$D = \left[\begin{array}{c|c|c} d_H I_m & & \\ \hline & d & \\ \hline & & d_h I_m \end{array} \right].$$

- Such a scaling **preserves** the **Toeplitz** structure and the ***M*-matrix** property of the submatrices.
- Analysis depends on these properties and on ratios between elements in the same row such as $|b/a|$ and $|b_H/c_H|$. These **ratios are invariant** under diagonal scaling.
- Consequently, all **convergence bounds hold** for the multiplicative Schwarz method applied to $DAx = Db$.

Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?
- 3 Multiplicative Schwarz method
- 4 Convergence analysis
- 5 Schwarz method as a preconditioner**
- 6 Numerical examples

Schwarz method as a preconditioner

We have consistent scheme

$$x^{(k+1)} = T x^{(k)} + v.$$

Hence, x solves $Ax = b$ and also “the **preconditioned** system”

$$(I - T)x = v.$$

Schwarz method as a preconditioner

We have consistent scheme

$$x^{(k+1)} = T x^{(k)} + v.$$

Hence, x solves $Ax = b$ and also “the **preconditioned** system”

$$(I - T)x = v.$$

We can formally define a **preconditioner** M such

$$Ax = b \Leftrightarrow M^{-1}Ax = M^{-1}b \Leftrightarrow (I - T)x = v.$$

Clearly $M = A(I - T)^{-1}$.

Schwarz method as a preconditioner

We have consistent scheme

$$x^{(k+1)} = T x^{(k)} + v.$$

Hence, x solves $Ax = b$ and also “the **preconditioned** system”

$$(I - T)x = v.$$

We can formally define a **preconditioner** M such

$$Ax = b \Leftrightarrow M^{-1}Ax = M^{-1}b \Leftrightarrow (I - T)x = v.$$

Clearly $M = A(I - T)^{-1}$. Then

$$\begin{aligned}x^{(k+1)} &= x^{(k)} + (I - T)(x - x^{(k)}) \\ &= x^{(k)} + M^{-1}r^{(k)}.\end{aligned}$$

Schwarz method as a preconditioner

for GMRES

- The multiplicative **Schwarz** method as well as **GMRES** applied to the preconditioned system obtain their approximations from **the same Krylov subspace**.
- In terms of the residual norm, the preconditioned **GMRES** will **always** converge **faster** than the multiplicative Schwarz.

Schwarz method as a preconditioner

for GMRES

- The multiplicative **Schwarz** method as well as **GMRES** applied to the preconditioned system obtain their approximations from **the same Krylov subspace**.
- In terms of the residual norm, the preconditioned **GMRES** will **always** converge **faster** than the multiplicative Schwarz.
- Moreover, in this case, the iteration matrix T has **rank-one structure**, and

$$\dim(\mathcal{K}_k(I - T, r_0)) \leq 2.$$

- Therefore, GMRES converges in **at most 2 steps**,
... a motivation for more dimensional cases.

How to multiply by T

Schwarz or preconditioned GMRES \rightarrow only multiply by T ,

$$T = (I - P_2)(I - P_1), \quad P_i = R_i^T (R_i A R_i^T)^{-1} R_i A,$$

i.e., to solve systems of the form ($m = n - 1 = \frac{N}{2} - 1$)

$$\left[\begin{array}{c|c} A_H & \\ \hline & b_H \\ \hline c & a \end{array} \right] \begin{bmatrix} y_{1:m} \\ y_{m+1} \end{bmatrix} = \begin{bmatrix} z_{1:m} \\ z_{m+1} \end{bmatrix}.$$

How to multiply by T

Schwarz or preconditioned GMRES \rightarrow only multiply by T ,

$$T = (I - P_2)(I - P_1), \quad P_i = R_i^T (R_i A R_i^T)^{-1} R_i A,$$

i.e., to solve systems of the form ($m = n - 1 = \frac{N}{2} - 1$)

$$\left[\begin{array}{c|c} A_H & \\ \hline & b_H \\ \hline c & a \end{array} \right] \begin{bmatrix} y_{1:m} \\ y_{m+1} \end{bmatrix} = \begin{bmatrix} z_{1:m} \\ z_{m+1} \end{bmatrix}.$$

Using the **Schur complement**,

$$\left(A_H - \frac{b_H c}{a} e_m e_m^T \right) y_{1:m} = z_{1:m} - z_{m+1} \frac{b_H}{a} e_m.$$

Then apply **Sherman-Morrison** formula.

We need only to solve systems with A_H (Toeplitz)!

Outline

- 1 Shishkin mesh and discretization
- 2 How to solve the linear system?
- 3 Multiplicative Schwarz method
- 4 Convergence analysis
- 5 Schwarz method as a preconditioner
- 6 Numerical examples**

Numerical examples

Consider

$$-\epsilon u'' + u' = 1, \quad u(0) = 0, \quad u(1) = 0,$$

i.e.

$$\alpha = 1, \quad \beta = 0, \quad f(x) \equiv 1.$$

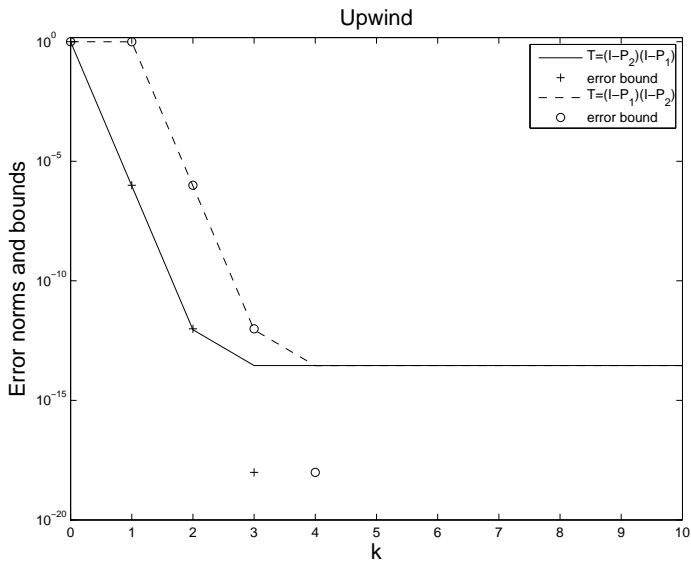
Choose $N = 198$, various values of ϵ .

| ϵ | upwind | | central differences | |
|------------|----------------------|----------------------|----------------------|----------------------|
| | ρ_{up} | our bound | ρ_{cd} | our bound |
| 10^{-8} | 9.4×10^{-7} | 9.9×10^{-7} | 1.8×10^{-4} | 3.9×10^{-4} |
| 10^{-6} | 9.4×10^{-5} | 9.9×10^{-5} | 1.8×10^{-2} | 3.9×10^{-2} |
| 10^{-4} | 9.3×10^{-3} | 9.8×10^{-3} | 8.3×10^{-1} | 3.8×10^{-0} |

$$\rho_{up} < \frac{\epsilon}{\epsilon + \alpha H}, \quad \rho_{cd} < \frac{2m\epsilon}{\epsilon + \frac{\alpha H}{2}}.$$

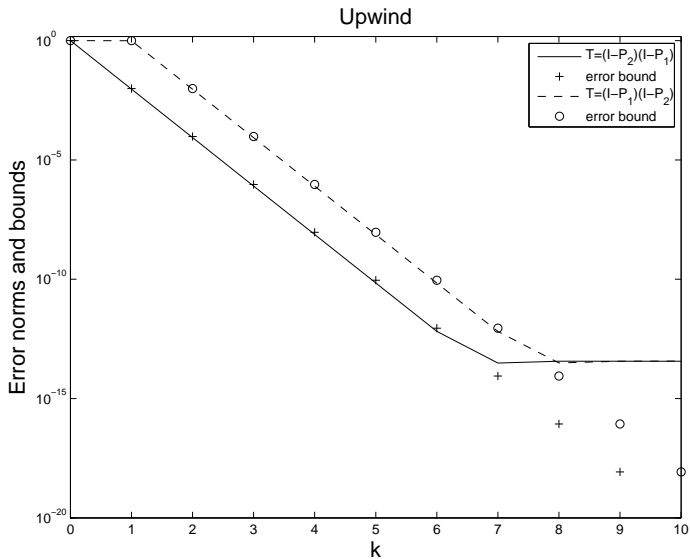
Numerical examples

Upwind, $\epsilon = 10^{-8}$



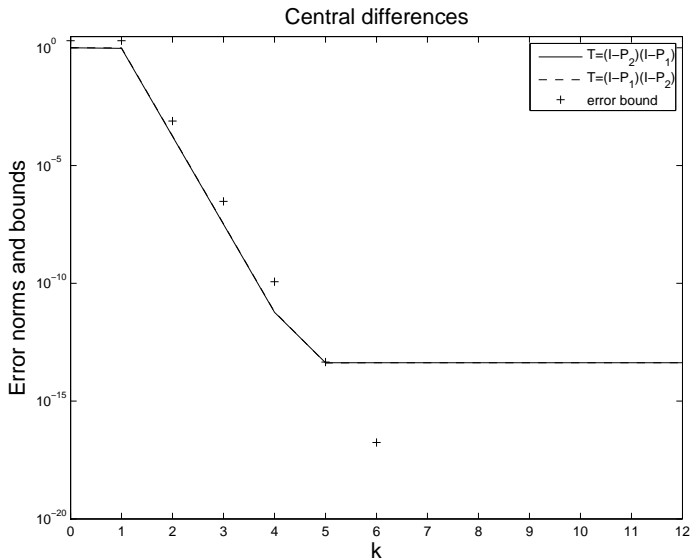
Numerical examples

Upwind, $\epsilon = 10^{-4}$



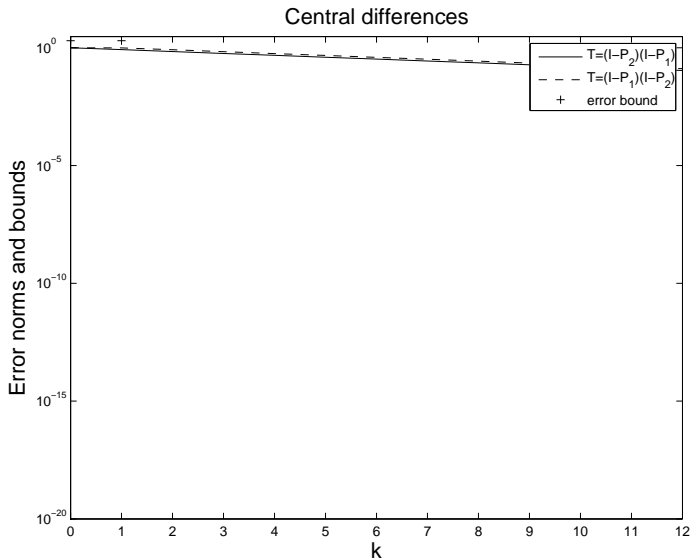
Numerical examples

Central differences, $\epsilon = 10^{-8}$



Numerical examples

Central differences, $\epsilon = 10^{-4}$



Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.

Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.
- The **matrices** that arise from the upwind and the central difference schemes are nonsymmetric and **highly nonnormal**.

Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.
- The **matrices** that arise from the upwind and the central difference schemes are nonsymmetric and **highly nonnormal**.
- For the **upwind scheme**, we proved **rapid convergence** of the multiplicative Schwarz method in the most relevant case $N\epsilon < \alpha$.

Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.
- The **matrices** that arise from the upwind and the central difference schemes are nonsymmetric and **highly nonnormal**.
- For the **upwind scheme**, we proved **rapid convergence** of the multiplicative Schwarz method in the most relevant case $N\epsilon < \alpha$.
- The convergence for the **central difference** scheme is **slower**, but still rapid, when $N^2\epsilon < \alpha$ and if $N/2 - 1$ is even.

Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.
- The **matrices** that arise from the upwind and the central difference schemes are nonsymmetric and **highly nonnormal**.
- For the **upwind scheme**, we proved **rapid convergence** of the multiplicative Schwarz method in the most relevant case $N\epsilon < \alpha$.
- The convergence for the **central difference** scheme is **slower**, but still rapid, when $N^2\epsilon < \alpha$ and if $N/2 - 1$ is even.
- Thanks to the **rank-one structure** of T , the preconditioned GMRES converges in **two steps**.

Conclusions and further work

- We considered finite difference discretizations of the 1D singularly-perturbed convection-diffusion equation posed on a **Shishkin mesh**.
- The **matrices** that arise from the upwind and the central difference schemes are nonsymmetric and **highly nonnormal**.
- For the **upwind scheme**, we proved **rapid convergence** of the multiplicative Schwarz method in the most relevant case $N\epsilon < \alpha$.
- The convergence for the **central difference** scheme is **slower**, but still rapid, when $N^2\epsilon < \alpha$ and if $N/2 - 1$ is even.
- Thanks to the **rank-one structure** of T , the preconditioned GMRES converges in **two steps**.
- Inspired by 1D case (**preconditioner**, **low-rank** structure), we can study **2D case**.

Related papers

- C. Echeverría, J. Liesen, P. Tichý, and D. Szyld, [Convergence of the multiplicative Schwarz method for singularly perturbed convection-diffusion problems discretized on a Shishkin mesh, (2016), in preparation]
- J. Miller, E. O'Riordan, and G. Shishkin, [Fitted Numerical Methods for Singular Perturbation Problems: Error Estimates in the Maximum Norm for Linear Problems in One and Two Dimensions, World Scientific, 1996.]
- H-G. Roos, M. Stynes, L. Tobiska, [Robust Numerical Methods for Singularly Perturbed Differential Equations, second edition, Springer Series in Computational Mathematics, Springer-Verlag, Berlin, 2008, 604 pp.]
- M. Stynes, [Steady-state convection-diffusion problems, Acta Numerica, 14 (2005), pp. 445–508.]

Thank you for your attention!