PRECONDITIONING THE HELMHOLTZ EQUATION WITH THE SHIFTED LAPLACIAN AND FABER POLYNOMIALS*

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Abstract. We introduce a new polynomial preconditioner for solving the discretized Helmholtz equation preconditioned with the complex shifted Laplace (CSL) operator. We exploit the localization of the spectrum of the CSL-preconditioned system to approximately enclose the eigenvalues by a non-convex 'bratwurst' set. On this set, we expand the function 1/z into a Faber series. Truncating the series gives a polynomial, which we apply to the Helmholtz matrix preconditioner by the shifted Laplacian to obtain a new preconditioner, the Faber preconditioner. We prove that the Faber preconditioner is nonsingular for degrees one and two of the truncated series. Our numerical experiments (for problems with constant and varying wavenumber) show that the Faber preconditioner reduces the number of GMRES iterations.

Key words. Helmholtz equation, shifted Laplace preconditioner, iterative methods, GMRES, preconditioning, Faber polynomials, 'bratwurst' sets

AMS subject classifications. 65F08, 65F10, 30C10, 30C20

1. Introduction. We study the solution of linear systems of equations

arising from the discretization of the Helmholtz equation. The Helmholtz (or reduced wave) equation results from the modelling of time-harmonic solutions of the standing wave equation. Such models appear in various branches of science and engineering, for example, in acoustics, seismic exploration, and medical imaging. Discretizing the Helmholtz equation on a domain equipped with boundary conditions by the finite difference or finite element method yields the linear system (1.1) with a matrix *A* that is complex symmetric, non-Hermitian, and indefinite. In practical applications, the system of equations is too large to be solved by direct methods, and iterative solvers are used instead. However, most classical iterative methods are known to perform poorly when applied to the Helmholtz equation; see the survey [22].

A standard approach is to use Krylov subspace methods in combination with a preconditioner, that is, instead of solving (1.1) directly, one finds a matrix M^{-1} that approximates A^{-1} and solves the system

$$AM^{-1}y = b$$

with a Krylov method and computes $x = M^{-1}y$. In the last two decades, many different preconditioners for the Helmholtz problem have been described. Some of the most efficient ones are domain-decomposition-based preconditioners (in particular sweeping-type preconditioners) and the shifted Laplacian. These two techniques are also used in combination; see [16, 24, 27, 53, 59] and the survey articles [17, 22, 23] for more references.

Among the latter, a widely used method is the complex shifted Laplacian (CSL) introduced in [20] and further developed in [19, 21]; see also [1, 4, 7, 23, 45]. In [18, 25, 49, 50], multilevel Krylov and multilevel deflation techniques to accelerate the CSL preconditioner are introduced and analyzed. For a survey, see [17], and for an overview of recent developments and applications of the CSL preconditioner, see the book [34]. The CSL preconditioner is

^{*}Received July 17, 2019. Accepted July 16, 2021. Published online on September 6, 2021. Recommended by Bernd Beckermann.

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one of the building blocks of some of the most efficient preconditioners for the Helmholtz equation currently in use; see [34]. In this work, we investigate the combination of the CSL preconditioner with a polynomial preconditioner.

Polynomial preconditioning for the solution of linear systems is a well-known technique; see [10, 35, 47] and the books [26, 48, 57]. There, the preconditioner is a polynomial in A, i.e., $M^{-1} = p(A)$. One idea to obtain such a polynomial is to expand A^{-1} into a convergent polynomial series and use a truncation of the series as preconditioner. The simplest form is a Neumann series expansion. Other approaches construct a polynomial by solving a minimization problem, e.g., Chebyshev polynomials [33] or least-squares polynomials [55], which both require an estimate of the spectrum of the matrix, or polynomials related to the minimum residual polynomial generated by running a few steps of a GMRES iteration [37]. Polynomial preconditioning for the Helmholtz equation was used in [8]. There, the inverse of the Helmholtz operator is expanded into a Taylor series around a complex shifted Laplacian operator, and the resulting truncated series is used as a preconditioner. The preconditioner based on a rational approximation for indefinite linear systems introduced in [58] has also been applied to the Helmholtz equation.

Polynomial preconditioning can also be applied to the preconditioned system, i.e., instead of solving

$$AM^{-1}x = b,$$

one solves the polynomially preconditioned system

$$AM^{-1}p(AM^{-1})y = b,$$

with a suitable polynomial p, and computes $x = p(AM^{-1})y$. Polynomial approximation techniques can also be combined with multigrid and multilevel methods, e.g., as smoothers for difficult problems where typical smoothers only give slow convergence or even divergence of the multilevel scheme [15]. A polynomial related to the best approximation of 1/x in the supremum norm on an interval has been applied to two-level methods in [33] for the case that the eigenvalues lie inside a real interval.

A general principle underlying many of these approaches is that if a polynomial p^* approximates the function f(z) = 1/z on a set that contains the spectrum of A (or of the preconditioned matrix AM^{-1}), then $p^*(A)$ (respectively, $p^*(AM^{-1})$) can be used as a preconditioner for the linear system. In this work we follow this principle and approximate the function 1/z on a set E containing the spectrum of AM^{-1} by Faber polynomials. Faber polynomials have been used previously in numerical analysis and in numerical linear algebra, e.g., in [3, 11, 13, 14, 29, 42, 43, 52].

We exploit the fact that the spectrum of the CSL-preconditioned system AM^{-1} lies in the disk $\{z \in \mathbb{C} : |z - 1/2| \le 1/2\}$ (see Theorem 2.1), which was first proved in [56], and we approximately enclose the spectrum by a 'bratwurst' set [31]. The function f(z) = 1/z is analytic on our inclusion set E and can be expanded into a convergent Faber series, which we determine explicitly. We then use the truncated Faber series of the preconditioned matrix as a new preconditioner for a Krylov subspace method. In our numerical experiments we truncate the series after at most three terms.

The polynomial preconditioner developed in [8] for the Helmholtz equation reduces the number of iterations but the overall CPU time increases. Our experiments show that our new polynomial preconditioner of the CSL preconditioner reduces the number of iterations and sometimes also the CPU time compared to the use of the CSL preconditioner only.

This paper is organized as follows. In Section 2 we introduce the discrete Helmholtz problem and the shifted Laplacian. The polynomial preconditioner is constructed in Section 3,

where we also briefly review the theory of Faber series, the definition of 'bratwurst' sets, and the use of these for polynomial preconditioning. Section 4 presents the results of numerical experiments for one- and two-dimensional Helmholtz problems, and in Section 5 we present some conclusions.

2. The Helmholtz equation and the complex shifted Laplace preconditioner. In this section we discuss the discretization of the Helmholtz equation and the complex shifted Laplace operator. For a bounded domain $\Omega \subset \mathbb{R}^d$ and $k \in \mathbb{R}$, the Helmholtz equation is given by

$$(2.1) \qquad \qquad -\Delta u - k^2 u = f,$$

with Dirichlet boundary conditions of the form

$$u = 0$$
 on $\partial \Omega$,

or first-order absorbing boundary conditions (sometimes also called impedance boundary conditions, e.g., in [23], or Sommerfeld condition, e.g., in [20])

$$\partial_n u - iku = 0$$
 on $\partial \Omega$.

The parameter k is the wavenumber. In many applications the dimension d is equal to 2 or 3, and the wavenumber can be large. After discretization by first-order finite differences on a grid of meshsize h or by linear finite elements, we obtain a discrete Helmholtz equation of the form

where

in which L is the discretization of the Laplacian, C arises from the discretization of the absorbing boundary conditions (so C = 0 in case of Dirichlet boundary conditions), and S results from the zeroth-order term [56]. The matrices L and S are real, symmetric, and positive definite, and C is positive semidefinite. Since the solutions to (2.1) are oscillatory, the number of gridpoints in one dimension must scale linearly with k. Therefore, the gridsize is chosen by requiring a fixed number of points per wavelength, typically at least 10 [2], or equivalently, keeping the quantity kh constant as k increases. This rule, however, is not sufficient to maintain the accuracy of the solutions at very high wavenumbers due to the 'pollution effect' caused by numerical dispersion [30]. It has been shown that for high wavenumbers, the gridsize should scale at least linearly with $k^{3/2}$ [41]. For some applications, this leads to very large linear systems of equations.

For $\varepsilon \in \mathbb{R}$, the *complex shifted Laplacian preconditioner* (CSL preconditioner) corresponds to the discretization of the problem

(2.4)
$$-\Delta u - (k^2 + i\varepsilon)u = f$$

with the same boundary conditions as the original Helmholtz problem. The choice of the boundary conditions can be justified by the results in [23, 39]. This results in the discrete problem $A_{\varepsilon}x = b$ with the coefficient matrix

(2.5)
$$A_{\varepsilon} = L + ikC - (k^2 + i\varepsilon)S = A - i\varepsilon S.$$



FIG. 2.1. The spectrum of AA_{ε}^{-1} with Dirichlet (left) and absorbing (right) boundary conditions on $\Omega = [0, 1]^2$ with parameters k = 15, $\beta = 0.5$.

The matrix A_{ε} is used as a preconditioner for (2.2). The resulting right-preconditioned problem is

$$AA_{\varepsilon}^{-1}y = b, \qquad x = A_{\varepsilon}^{-1}y.$$

The CSL problem in (2.4) can be interpreted physically as an artificially damped Helmholtz problem, where the damping allows the preconditioner A_{ε} to be inverted efficiently, for example, by standard multigrid methods. The shift should be chosen to minimize the number of iterations required to solve the preconditioned system, while still being able to (approximately) solve the shifted system. It has been shown in [23] that if the GMRES method is used for the outer Krylov iteration, then a sufficient condition to ensure a number of iterations independent of the wavenumber is that the shift scales linearly with respect to k, that is, $\varepsilon = O(k)$. This, however, poses difficulties for the inversion of the preconditioner A_{ε} , in particular when multigrid methods are used. Indeed, it was shown in [5] that the shift ε must be at least $O(k^2)$ for the shifted Laplace preconditioner to be able to be inverted using multigrid methods for high wavenumbers, and earlier publications indicate that the choice $\varepsilon = k^2\beta$ for a fixed $\beta \in (0, 1)$ can lead to an efficient preconditioner for moderate wavenumbers when A_{ε} is inverted with the multigrid method [19]. This has been further justified by a spectral analysis and local Fourier analysis in [7, 56], where the authors found that $\beta = 0.5$ leads to a nearly optimal method in terms of Krylov iterations. We use this choice in our numerical experiments.

The eigenvalues of the preconditioned matrix AA_{ε}^{-1} lie in a disk. This was first proved in [56]; see also [25] for an alternative proof. We use this localization of the spectrum of AA_{ε}^{-1} to develop our polynomial preconditioner.

THEOREM 2.1. In dimension d = 1, 2, or 3, let A and A_{ε} be given by (2.3) and (2.5), respectively, and let $\varepsilon = k^2 \beta$ with $\beta > 0$. Then the spectrum of the preconditioned matrix AA_{ε}^{-1} lies in the closed disk

(2.6)
$$\{z \in \mathbb{C} : |z - 1/2| \le 1/2\}.$$

In case of Dirichlet boundary conditions, the spectrum of AA_{ε}^{-1} lies on the boundary of (2.6).

Theorem 2.1 is illustrated in Figure 2.1 for Helmholtz problems on $\Omega = [0, 1] \times [0, 1]$ with k = 15 and $\beta = 0.5$.

3. Construction of the Faber preconditioner from the Faber series. In order to construct a polynomial preconditioner for solving the CSL-preconditioned Helmholtz equation

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FIG. 3.1. The set E and several contour lines Γ_r .

 $AA_{\varepsilon}^{-1}x = b$ by GMRES, we approximately enclose the spectrum of AA_{ε}^{-1} by a non-convex set and expand 1/z into a Faber series on the inclusion set. The truncated series is a polynomial. Applying it to AA_{ε}^{-1} gives the Faber preconditioner.

3.1. Faber polynomials and the Faber series. Faber polynomials allow the series expansion of analytic functions on certain compact subsets of \mathbb{C} . In this section, we summarize the definition and some properties of the Faber polynomials and the Faber series, which can be found alongside with further properties in, e.g., [9], [54, Ch. III], [40, Sect. 14], or [51, Ch. 2].

Let $E \subset \mathbb{C}$ be a compact set (not a single point) with simply connected complement $E^c := \widehat{\mathbb{C}} \setminus E$ in the extended complex plane $\widehat{\mathbb{C}} = \mathbb{C} \cup \{\infty\}$. By the Riemann mapping theorem there exists a unique conformal map

(3.1)
$$\psi: \{w \in \widehat{\mathbb{C}}: |w| > 1\} \to E^c \quad \text{with} \quad \psi(\infty) = \infty, \quad \psi'(\infty) > 0.$$

We denote the image of the circle |w| = r under ψ by

$$\Gamma_r = \psi(\{w \in \mathbb{C} : |w| = r\}), \quad r > 1$$

The curves Γ_r are the level curves of the Green function for E^c with the pole at infinity, given by $g_{E^c}(z) = \log |\psi^{-1}(z)|$. Figure 3.1 displays a compact set E (bounded by the solid line) and several curves Γ_r (the dashed lines).

For a fixed $z \in \mathbb{C}$, the function $\psi'(w)/(\psi(w) - z)$ is analytic in a neighbourhood of infinity and can thus be expanded into a Laurent series

$$\frac{\psi'(w)}{\psi(w)-z} = \sum_{k=0}^{\infty} \frac{F_k(z)}{w^{k+1}}, \qquad z \text{ interior to } \Gamma_r, \quad |w| > r.$$

Each coefficient $F_k(z)$ is a polynomial of degree k, the k-th Faber polynomial of E. The leading coefficient of $F_k(z)$ is $(\psi'(\infty))^{-k}$, where $\psi'(\infty) = c(E)$ is the logarithmic capacity of E.

Let f be an analytic function on the compact set E, that is, f is analytic in an open set containing E. Therefore, there exists a largest $\rho > 1$ such that f is analytic in the interior of the curve Γ_{ρ} but not in the interior of a larger curve (for an entire function, $\rho = \infty$). The

function f has a unique representation as a *Faber series*

(3.2)
$$f(z) = \sum_{k=0}^{\infty} a_k F_k(z)$$

which converges absolutely in the interior of Γ_{ρ} , uniformly on compact sets contained in the interior of Γ_{ρ} , and diverges in the exterior of Γ_{ρ} . The coefficients are given by

(3.3)
$$a_k = \frac{1}{2\pi i} \int_{|w|=r} \frac{f(\psi(w))}{w^{k+1}} \, dw,$$

where $1 < r < \rho$ is arbitrary. Here and in the sequel we integrate over the circle oriented as usual in the positive sense (counter-clockwise). In our application, we compute the Faber polynomials and the coefficients analytically.

EXAMPLE 3.1.

- 1. For the disk $E = \{z \in \mathbb{C} : |z| \le 1\}$, we have $\psi(z) = z$, $F_k(z) = z^k$, and (3.2) is a Taylor series with radius of convergence ρ .
- 2. For the interval E = [-1, 1], we have $\psi(z) = \frac{1}{2}(z+z^{-1})$, and the Faber polynomials are related to the Chebyshev polynomials of the first kind by $F_0(z) = T_0(z)$ and $F_k(z) = 2T_k(z)$, for $k \ge 1$.

The Faber series (3.2) converges geometrically to f on E, which is the fastest possible convergence in an n-th root sense. This is expressed by

$$\limsup_{n \to \infty} \|f - \sum_{k=0}^{n} a_k F_k\|_E^{1/n} = \frac{1}{\rho};$$

see [28]. Here $\|\cdot\|_E$ denotes the supremum norm on E. Estimates for $\|f - \sum_{k=0}^n a_k F_k\|_E$ for fixed n are given, e.g., in [14, 32], [51, p. 142], and [54, Ch. 8 and 9]. We give an error estimate for our case in Section 3.3.

Faber series can also be used to define matrix functions as follows. Let A be a square matrix whose eigenvalues lie in E. Inserting A in the scalar Faber series (3.2) yields

(3.4)
$$f(A) = \sum_{k=0}^{\infty} a_k F_k(A),$$

and this series of matrices converges in any matrix norm $\|\cdot\|$ with

$$\limsup_{n \to \infty} \|f(A) - \sum_{k=0}^n a_k F_k(A)\|^{1/n} \le \frac{1}{\rho};$$

see [28, p. 316] or [42], where functions of matrices are computed by (3.4). If the spectrum of A is contained in the interior of Γ_{ρ} but not necessarily in E, then the series still converges but at a lower rate.

3.2. The inclusion set. We construct a set containing (most of) the spectrum of the CSL-preconditioned Helmholtz operator AA_{ε}^{-1} (see Figure 2.1) and excluding the origin. The inclusion set is a scaled and shifted variant of the 'bratwurst' sets introduced in [31] and has the form depicted in Figure 3.1. The 'bratwurst' sets are defined through a conformal map



FIG. 3.2. The sets $\Omega(1, \pi/2, \sigma)$ for $\sigma = 1$ (arc), $\sigma = 1.1$, and $\sigma = 1.15$.

from the exterior of the unit disk onto the exterior of the set. The following theorem describes their construction.

THEOREM 3.2 ([31, Theorem 3.1]). Let $\lambda \in \mathbb{C}$ with $|\lambda| = 1$, $\phi \in [0, 2\pi[$, and $\sigma \in [1, P[$, where $P = \tan(\phi/4) + \frac{1}{\cos(\phi/4)}$, and let

$$\widetilde{\psi}(w) = \frac{(w - \lambda N)(w - \lambda M)}{(N - M)w + \lambda(NM - 1)}$$

where $M = \frac{\sigma^2 - 1}{2\sigma \tan(\phi/4)}$ and $N = \frac{1}{2} \left(\frac{P}{\sigma} + \frac{\sigma}{P}\right)$. Define $\Omega = \Omega(\lambda, \phi, \sigma)$ to be the compact set bounded by $\widetilde{\psi}(\{w \in \mathbb{C} : |w| = 1\})$. Then $\widetilde{\psi}$ is a bijective conformal map from the exterior of the unit circle onto $\widehat{\mathbb{C}} \setminus \Omega$ that satisfies $\widetilde{\psi}(\infty) = \infty$ and $\psi'(\infty) > 0$. We further have $\lambda \notin \Omega$ and $\{\lambda e^{i\beta} : \phi/2 \le \beta \le 2\pi - \phi/2\} \subseteq \Omega$.

The constants M and N satisfy $0 \le M < 1 < N$; see [31, Eq. (3.16)].

Figure 3.2 displays the sets $\Omega(1, \pi/2, \sigma)$ for different σ . The parameter λ indicates where the "opening" is, ϕ is related to the width of the opening, and σ to how "thick" Ω is. For $\sigma = 1$ it is a Jordan arc [31, Example 3.2] and for $1 < \sigma < P$ a set bounded by an analytic Jordan curve.

The eigenvalues of AA_{ε}^{-1} lie in the disk $\{z \in \mathbb{C} : |z - \frac{1}{2}| \le \frac{1}{2}\}$. We require an inclusion set of the eigenvalues that excludes the origin, which is the pole of 1/z. Let

(3.5)
$$E = \frac{1}{2}\Omega(-1,\phi,\sigma) + \frac{1}{2}.$$

The choice $\lambda = -1$ implies that $-1 \notin \Omega(-1, \phi, \sigma)$, and thus $0 \notin E$. The parameters ϕ and σ will be chosen in Section 3.7. The function

(3.6)
$$\psi(w) = \frac{\psi(w) + 1}{2} = \frac{w^2 + 2Nw + 1}{2(N - M)w + 2(1 - MN)}$$

is the exterior mapping function of E in (3.5), i.e., ψ is conformal and satisfies (3.1).

3.3. The Faber series of 1/z on the inclusion set. The Faber polynomials for the set E in (3.5) can be computed efficiently by a three-term recurrence; see [36, Section 3]. This follows from the fact that ψ in (3.6) is a rational exterior mapping function of type (2, 1). The Faber polynomials of E are given by $F_0(z) = 1$ and

(3.7)
$$F_n(z) = \widehat{F}_n(z) - \left(\frac{MN-1}{N-M}\right)^n, \quad n \ge 1,$$

where the *shifted Faber polynomials* \hat{F}_n satisfy the three-term recurrence

$$F_0(z) = 2,$$

$$\widehat{F}_1(z) = 2(N-M)z - 2N,$$

$$\widehat{F}_n(z) = (2(N-M)z - 2N)\widehat{F}_{n-1}(z) + (2(1-MN)z - 1)\widehat{F}_{n-2}(z), \quad n \ge 2.$$

In particular, the leading coefficient of $F_n(z)$ is $(2(N-M))^n$.

Since f(z) = 1/z is analytic on E (recall that $0 \notin E$), it has a convergent Faber series, whose coefficients can be computed explicitly. Figure 3.1 displays E and the contour Γ_{ρ} from the next theorem.

THEOREM 3.3. With the notation of Theorem 3.2, the function f(z) = 1/z is analytic on the set $E = \frac{1}{2}\Omega(-1, \phi, \sigma) + \frac{1}{2}$. Its Faber series

(3.8)
$$f(z) = \frac{1}{z} = \sum_{k=0}^{\infty} a_k F_k(z)$$

converges absolutely and locally uniformly interior to Γ_{ρ} and diverges exterior to Γ_{ρ} , where $\rho = N + \sqrt{N^2 - 1} > 1$. The coefficients are given by

(3.9)
$$a_k = (-1)^k \frac{\rho - M}{\rho^{k+1}}.$$

Proof. The existence and convergence properties of the Faber series of f follow from the general theory of Faber series; see Section 3.1. It remains to determine the number ρ and to show that the coefficients are given by (3.9).

Since f is analytic in \mathbb{C} except at z = 0, where it has a pole, the number ρ is determined by $0 \in \Gamma_{\rho}$, i.e., $\rho = |\psi^{-1}(0)| > 1$, where ψ is the conformal map from (3.6). The zeros of ψ are

(3.10)
$$\alpha := -N - \sqrt{N^2 - 1}$$
 and $\beta := -N + \sqrt{N^2 - 1}$.

In particular, $\alpha + \beta = -2N$ and $\alpha\beta = 1$. Since N > 1, we find

$$\alpha = -N - \sqrt{N^2 - 1} < -1 < \beta = -N + \sqrt{N^2 - 1} < 0.$$

Since ψ is defined in |w|>1, we have $\psi^{-1}(0)=-N-\sqrt{N^2-1}$ and

$$\rho = |\psi^{-1}(0)| = N + \sqrt{N^2 - 1} > 1.$$

Next, we compute the coefficients a_k . By (3.3), they are given by

(3.11)
$$a_k = \frac{1}{2\pi i} \int_{|t|=r} \frac{1}{\psi(t)} \frac{1}{t^{k+1}} dt$$

for any $1 < r < \rho$. The integrand is rational with poles at 0, α , and β , and we compute the integral (3.11) by the residue theorem. Note that α lies exterior to |t| = r since $|\alpha| = \rho > r$, and β lies interior to |t| = r since $|\beta| < 1 < r$. The residue at the simple pole at $t = \beta$ is given by

$$\operatorname{res}\left(\frac{1}{\psi(t)}\frac{1}{t^{k+1}},\beta\right) = \lim_{t \to \beta} \frac{t-\beta}{\psi(t)t^{k+1}} = \frac{2(N-M)\beta + 2(1-MN)}{\beta - \alpha}\frac{1}{\beta^{k+1}} \\ = \frac{N-M-\sqrt{N^2-1}}{\beta^{k+1}} = -\frac{M+\beta}{\beta^{k+1}},$$

where we inserted the values of α and β from (3.10). To compute the residue at t = 0, which is a pole of order k + 1, we consider the partial fraction decomposition

$$\frac{1}{\psi(t)} = \frac{2(N-M)t + 2(1-MN)}{t^2 + 2Nt + 1} = \frac{M+\alpha}{\alpha - t} + \frac{M+\beta}{\beta - t},$$

with α and β from (3.10). For sufficiently small |t| we have

 \Box

$$\frac{1}{\psi(t)} = \frac{M+\alpha}{\alpha} \frac{1}{1-t/\alpha} + \frac{M+\beta}{\beta} \frac{1}{1-t/\beta} = \frac{M+\alpha}{\alpha} \sum_{j=0}^{\infty} \frac{t^j}{\alpha^j} + \frac{M+\beta}{\beta} \sum_{j=0}^{\infty} \frac{t^j}{\beta^j},$$

so that

$$\operatorname{res}\left(\frac{1}{\psi(t)}\frac{1}{t^{k+1}},0\right) = \frac{M+\alpha}{\alpha^{k+1}} + \frac{M+\beta}{\beta^{k+1}}.$$

Together, the residue theorem yields for (3.11)

$$a_k = \frac{M+\alpha}{\alpha^{k+1}} = (-1)^k \frac{N+\sqrt{N^2-1}-M}{(N+\sqrt{N^2-1})^{k+1}}, \qquad k \ge 0,$$

which concludes the proof.

The quantity $\frac{1}{|\psi^{-1}(0)|} = \frac{1}{\rho}$ is known as the *asymptotic convergence factor* for polynomials from $\mathcal{P}_n(0) = \{p : p \text{ polynomial with } \deg(p) \le n, p(0) = 1\}$ associated with E; see [12, 13] and [31, Eq. (2.6)].

3.4. Rate of convergence of the Faber series. How well does the truncated Faber series approximate 1/z in Theorem 3.3? Let $s_n = \sum_{k=0}^n a_k F_k$ be a partial sum of the Faber series (3.8). Then

$$||f - s_n||_E = ||\sum_{k=n+1}^{\infty} a_k F_k||_E \le \sum_{k=n+1}^{\infty} |a_k|||F_k||_E.$$

We have $|a_k| = (\rho - M)/\rho^{k+1}$ from (3.9) and $||F_k||_E \leq V/\pi$, where V is the *boundary rotation* of E; see [54, p. 182]. We give the definition of the boundary rotation and discuss its computation in Appendix A. The approximation error thus satisfies

$$\|f - s_n\|_E \le \sum_{k=n+1}^{\infty} \frac{\rho - M}{\rho^{k+1}} \frac{V}{\pi} = \frac{V}{\pi} \frac{\rho - M}{\rho - 1} \frac{1}{\rho^{n+1}}$$

If the origin (the pole of 1/z) is close to E, then we expect that $\rho > 1$ is close to 1. For instance, for $E = \frac{1}{2}\Omega(-1, \pi/2, 1.1) + \frac{1}{2}$, we have $||f - s_n||_E \le 14.31/1.36^{n+1}$.

In theory, we can solve the CSL-preconditioned Helmholtz equation

by computing

$$x = (AA_{\varepsilon}^{-1})^{-1}b = f(AA_{\varepsilon}^{-1})b = \sum_{k=0}^{\infty} a_k F_k(AA_{\varepsilon}^{-1})b$$

with the Faber series (3.8) of 1/z. The series of matrices converges provided that the spectrum of AA_{ε}^{-1} is contained in E (or in the interior of the curve Γ_{ρ}). For practical purposes, however,

the rate of convergence of the series is too low. This is due to the expected slow convergence of the scalar Faber series and also to some eigenvalues that are close to the origin, which further decreases the rate of convergence of the series. Therefore, instead of solving (3.12) directly with the Faber series, we take a low-degree partial sum of the Faber series as a polynomial preconditioner for GMRES.

3.5. The Faber preconditioner. We construct our polynomial preconditioner for solving (3.12) with GMRES by truncating the Faber series of 1/z. More precisely, we consider the polynomially preconditioned system

where the polynomial $s_n(z) = \sum_{k=0}^n a_k F_k(z)$ is the *n*-th partial sum of the Faber series (3.8) of 1/z, and $n \ge 1$. We call the polynomial preconditioner $s_n(AA_{\varepsilon}^{-1})$ the Faber preconditioner of degree *n*, denoted by FP(*n*).

When a Krylov method is used to solve (3.13), every iteration requires $deg(s_n)+1 = n+1$ matrix-vector products with AA_{ε}^{-1} , that is, n + 1 approximate inversions of A_{ε} . This limits our choice to small degrees n, and in our experiments we consider $n \leq 3$.

We also have to ensure that the Faber preconditioner $s_n(AA_{\varepsilon}^{-1})$ is nonsingular. In the limit $n \to \infty$ we have

$$\lim_{n \to \infty} s_n(AA_{\varepsilon}^{-1}) = f(AA_{\varepsilon}^{-1}) = (AA_{\varepsilon}^{-1})^{-1},$$

which guarantees that $s_n(AA_{\varepsilon}^{-1})$ is nonsingular for sufficiently large n. To show that the Faber preconditioner is nonsingular for small n requires more specific information on the location of the zeros of the polynomials s_n . We investigate this issue for n = 1, 2 in the following proposition. The proof is postponed to Appendix B.

PROPOSITION 3.4. Let $E = \frac{1}{2}\Omega(-1, \phi, \sigma) + \frac{1}{2}$ be a scaled and shifted 'bratwurst' set as in Theorem 3.2, and let $s_n(z)$ be the truncated Faber series of 1/z of degree n. If AA_{ε}^{-1} is a Helmholtz matrix preconditioned by the complex shifted Laplacian and n = 1, 2, then the Faber preconditioner $s_n(AA_{\varepsilon}^{-1})$ is nonsingular.

3.6. Evaluation of the Faber preconditioner. In our numerical experiments, we compute $s_n(B)v$ for $B = AA_{\varepsilon}^{-1}$ and a vector v by using the three-term recurrence of the shifted Faber polynomials. We have

$$s_n(z) = \sum_{k=0}^n a_k F_k(z) = \sum_{k=0}^n a_k \widehat{F}_j(z) - \sum_{k=0}^n a_k \left(\frac{MN-1}{N-M}\right)^k.$$

Thus, $s_n(z)$ can be computed iteratively by $\hat{s}_0(z)$, which collects all constant terms, and $\hat{s}_m(z) = \hat{s}_{m-1}(z) + a_m \hat{F}_m(z)$, which yields $\hat{s}_n(z) = s_n(z)$. In the process, the shifted Faber polynomials are computed with their three-term recurrence (see Section 3.3). Inserting the matrix and vector gives a short recurrence for computing $s_n(AA_{\varepsilon}^{-1})v$.

3.7. Determining parameters for the inclusion set. To determine suitable parameters ϕ and σ of the inclusion set (3.5), we compared the number of iterations and the computation time to solve

$$AA_{\varepsilon}^{-1}s_2(AA_{\varepsilon}^{-1})x = b$$

with (full) GMRES for the angles $\phi = 0.1\pi, 0.2\pi, \dots, 0.9\pi$ and the corresponding values $\sigma = 1:0.05: P(\phi)$ in the model problem 1 below. The number of iterations and times were the

smallest for σ close to 1. The effect of the angle seems to be less important. We therefore use $\phi = \frac{\pi}{2}$ and $\sigma = 1.005$ in our numerical experiments.

We give a first heuristic explanation. The rate of convergence of the Faber series of 1/zon E depends on the number ρ , which measures the distance of E to z = 0, i.e., the singularity of 1/z, with the level curves Γ_r ; recall Section 3.1. For fixed ϕ this distance is the largest for the arc and decreases as σ increases. Thus considering for E just an arc gives the fastest approximation of 1/z on E. Note that the Faber series also converges interior to other level curves Γ_r , $1 < r < \rho$, with the convergence being slower for larger r. Thus, even if the spectrum of the matrix AA_{ε}^{-1} is not contained in E but only in the interior of some level curve Γ_r with small r, we will have a high rate of convergence. A rigorous explanation for the matrix case is, however, subject to further work.

4. Numerical experiments. We present the results of numerical experiments for solving Helmholtz problems preconditioned with the complex shifted Laplacian in combination with Faber preconditioners. We consider four model problems: The Helmholtz equation

- 1. on (0,1) with Dirichlet boundary conditions,
- 2. on (0,1) with absorbing boundary conditions,
- 3. on $(0,1)^2$ with absorbing boundary conditions,

4. on $(0,1)^2$ with absorbing boundary conditions and variable wavenumber. We solve the linear system of algebraic equations

using the Faber preconditioner as described in Section 3.5, i.e., we solve

$$AA_{\varepsilon}^{-1}s_n(AA_{\varepsilon}^{-1})y = b$$

with full or restarted GMRES, and then compute $x = s_n(AA_{\varepsilon}^{-1})y$. Here A_{ε}^{-1} is the complex shifted Laplace preconditioner with $\varepsilon = 0.5k^2$ (see Section 2), and s_n is the truncated Faber series of 1/z of degree n on the inclusion set (3.5) with parameters $\phi = \pi/2$ and $\sigma = 1.005$. Then the Faber preconditioners $s_1(AA_{\varepsilon}^{-1})$ and $s_2(AA_{\varepsilon}^{-1})$ are nonsingular (Proposition 3.4), and also $s_3(AA_{\varepsilon}^{-1})$ is nonsingular since the numerically computed roots of s_3 lie outside the disk $\{z \in \mathbb{C} : |z - 1/2| \le 1/2\}$. For comparison, we also solve the system (4.1) preconditioned with the CSL but without the Faber preconditioner.

In our experiments, we consider two variants to invert the preconditioner A_{ε}^{-1} : via a precomputed LU decomposition in the method denoted by CSL(LU) (only for 1D problems and for reference, since this is not feasible for practical problems) and with a V(1, 1) (or F(1, 1) for 2D problems) multigrid cycle with ω -Jacobi as a smoother (with relaxation parameter $\omega = 2/3$) in the method CSL(MG). The methods FP(n)+CSL(LU), and FP(n)+CSL(MG) result from combining the respective variant of the shifted Laplacian with a Faber preconditioner of degree n. Since the preconditioner is relatively expensive to compute, we only use low degrees n = 1, 2, 3.

All experiments have been performed in MATLAB 2018b on a Laptop with an i7 processor and 16 GB RAM. We use our own discretization of the Helmholtz equation and multigrid implementation.

4.1. Model problem 1. We consider a Helmholtz problem in one dimension with Dirichlet boundary conditions:

$$-u'' - k^2 u = f$$
 in (0, 1),
 $u(0) = u(1) = 0$,

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TABLE 4.1 Model problem 1: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with an LU factorization.

$_{k}$	CSL(LU)			FP(1)+CSL(LU)			FP(2)+CSL(LU)			FP(3)+CSL(LU)		
	Iter.	MV	Time(s)	Iter.	MV	MV Time(s)		MV	Time(s)	Iter.	MV	Time(s)
20	16	16	0.059	17	34	0.020	13	39	0.045	11	44	0.048
40	25	25	0.037	22	44	0.012	16	48	0.045	14	56	0.046
60	32	32	0.044	26	52	0.026	19	57	0.050	16	64	0.050
80	38	38	0.045	28	56	0.029	21	63	0.052	17	68	0.050
100	47	47	0.053	33	66	0.022	25	75	0.058	21	84	0.054
120	51	51	0.068	35	70	0.034	25	75	0.060	21	84	0.058
150	61	61	0.078	40	80	0.042	30	90	0.078	25	100	0.062
200	76	76	0.112	49	98	0.040	36	108	0.088	29	116	0.078
400	131	131	0.309	79	158	0.153	57	171	0.132	47	188	0.133
600	182	182	0.717	107	214	0.300	78	234	0.290	63	252	0.247
800	236	236	1.120	137	274	0.412	100	300	0.395	81	324	0.333
300	→ CS ×- ·FP → ·FP	L(LU) (1)+CS (2)+CS (3)+CS	SL(LU) SL(LU) SL(LU)			1 0 0 0 0 0 0		- CSL(I · FP(1) · FP(2) · FP(3)	JU) +CSL(LU) +CSL(LU) +CSL(LU)		ø	

FIG. 4.1. Model problem 1: Number of GMRES iterations (left) and time elapsed (right) for computing the shifted Laplacian with an LU factorization.

Wavenumber

where f is a point source centered at x = 1/2. We discretize this problem with second-order central differences on a uniform grid with 15 points per wavelength. This simple problem can be solved with a direct method, but we present it for reference.

First, we apply the Faber preconditioner of degree 1, 2, and 3 to accelerate full GMRES. The GMRES iteration is stopped when the relative residual is below 10^{-8} . Table 4.1 and Figure 4.1 provide the number of GMRES iterations, the number of matrix-vector products (MV), and the time elapsed (in seconds) when the shifted Laplacian is computed using forward and backward substitution from a precomputed LU factorization (only for reference, since this is not done for practical problems). The timings include the computation of the LU factorization of A_{ε} . For comparison, we repeat the same experiment with the shifted Laplace preconditioner computed approximately with a multigrid V(1, 1)-cycle and a Jacobi smoother with relaxation parameter $\omega = 2/3$ and display the results in Table 4.2 and Figure 4.2.

In both experiments, we see a reduction in the number of iterations when the CSL is augmented with the Faber preconditioner. This reduction is larger for the Faber preconditioner of higher degrees when the LU factorization is used but not when the shifted Laplacian is approximately computed with a multigrid cycle, for which there is not a significant difference in the number of iterations for the Faber preconditioner of degrees 1, 2, and 3.

TABLE 4.2 Model problem 1: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with a multigrid cycle.

k		CSL(MG) FP(1)+CSL(MG)				L(MG)	FP	(2)+CS	L(MG)	FP(3)+CSL(MG)		
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)
20	17	17	0.059	16	32	0.099	13	39	0.096	12	48	0.103
40	27	27	0.047	22	44	0.101	19	57	0.109	16	64	0.124
60	36	36	0.065	30	60	0.150	27	81	0.148	22	88	0.147
80	42	42	0.077	31	62	0.132	27	81	0.146	24	96	0.154
100	52	52	0.110	51	102	0.194	44	132	0.196	41	164	0.210
120	61	61	0.148	43	86	0.177	42	126	0.224	34	136	0.235
150	71	71	0.166	49	98	0.214	40	120	0.200	37	148	0.246
200	90	90	0.205	75	150	0.305	65	195	0.352	64	256	0.410
400	163	163	0.530	120	240	0.680	107	321	0.792	109	436	0.948
600	238	238	1.379	151	302	1.159	115	345	1.085	100	400	1.145
800	305	305	2.069	210	420	1.679	186	558	1.765	198	792	2.144
300	CS 	200	SL(MG) SL(MG) SL(MG)	500	600 70			- CSL(N - FP(1) - FP(2) - FP(3)	MG) +CSL(MG +CSL(MG +CSL(MG		0 600	700 800
5		200	Wavenumb	er	/(5 1	20	Wave	numher		, 000

FIG. 4.2. Model problem 1: Number of GMRES iterations (left) and time elapsed (right) for computing the shifted Laplacian with a multigrid cycle.

When full GMRES is used, one cannot expect that the number of matrix-vector products is reduced, nevertheless we observe some speedup when the LU factorization is used since the extra cost of applying the preconditioner per iteration is negligible and the number of operations for orthogonalization is reduced. When the preconditioner is approximately computed via the multigrid method, the application of the preconditioner is a more expensive operation, and the reduction in the number of iterations is not as large as when the CSL is computed with backward and forward substitution. Therefore there is no speedup.

4.2. Model problem 2. We consider a Helmholtz problem in one dimension with absorbing boundary conditions:

$$-u'' - k^2 u = f \quad \text{in } (0, 1),$$

$$u'(0) + iku(0) = 0,$$

$$u'(1) - iku(1) = 0,$$

where f is a point source centered at x = 1/2. We discretize this problem on a uniform grid with piecewise linear finite elements and 15 points per wavelength.

As in model problem 1, we apply the Faber preconditioner to accelerate full GMRES. The results are very similar to the previous experiment. We observe a decrease in the number

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TABLE 4.3 Model problem 2: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with an LU factorization.

k	CSL(LU) FP(1)+CSL(LU)				L(LU)	FF	P(2)+CS	L(LU)	FP(3)+CSL(LU)			
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)
20	18	18	0.036	19	38	0.011	16	48	0.044	13	52	0.049
40	28	28	0.041	25	50	0.014	19	57	0.047	16	64	0.048
60	36	36	0.045	29	58	0.028	22	66	0.054	18	72	0.053
80	42	42	0.065	31	62	0.031	23	69	0.053	20	80	0.054
100	50	50	0.057	36	72	0.030	27	81	0.068	23	92	0.056
120	55	55	0.072	38	76	0.036	28	84	0.082	24	96	0.063
150	65	65	0.087	43	86	0.044	32	96	0.076	27	108	0.066
200	80	80	0.111	51	102	0.041	38	114	0.086	31	124	0.087
400	135	135	0.313	81	162	0.153	59	177	0.144	48	192	0.137
600	188	188	1.051	111	222	0.439	80	240	0.366	65	260	0.312
800	241	241	1.624	140	280	0.624	102	306	0.505	83	332	0.419
300 - Store and Store and		P(1)+C(2)+C(2)+C(3)+C(3)+C(3)+C(3)+C(3)+C(3)+C(3)+C(3	SL(LU) SL(LU) SL(LU)	500	600 70			- CSL(I • FP(1) • FP(2) • FP(3)	LU) +CSL(LU) +CSL(LU) +CSL(LU)		8	700 800
			Wavenumb	er					Wave	number		

FIG. 4.3. Model problem 2: Number of GMRES iterations (left) and time elapsed (right) for computing the shifted Laplacian with an LU factorization.

of iterations which results in a speedup (the computation time is typically halved for degree n = 1, while the computation time is reduced by more than a half for higher wavenumbers for degrees 2 and 3) when the preconditioner is inverted using forward and backward substitution with a precomputed LU decomposition; see Table 4.3 and Figure 4.3. When a multigrid cycle is used, the decrease in the number of iterations is not as significant, and it does not translate to a faster iteration; see Table 4.4 and Figure 4.4.

4.3. Model problem 3. We consider a Helmholtz problem in two dimensions with absorbing boundary conditions:

$$\begin{aligned} -\Delta u - k^2 u &= f & \text{ in } \Omega = (0,1)^2, \\ \partial_n u - iku &= 0 & \text{ in } \partial \Omega, \end{aligned}$$

where f is a point source centered at (1/2, 1/2). This problem is discretized with second-order finite differences on a uniform grid with 12 points per wavelength [46].

We solve this problem with restarted and full GMRES. We begin with restarted GMRES and set the length of an outer cycle equal to 20 (i.e., we use GMRES(20)), and the wavenumber is k = 50. Here we set the degree of the Faber preconditioner to be equal to 1, 2, or 3 and plot the norm of the residual as a function of the number of matrix-vector products and use an

TABLE 4.4 Model problem 2: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with a multigrid cycle.

k	CSL(MG) FP((1)+CS	L(MG)	FP(2)+CSL(MG)			FP(3)+CSL(MG)			
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)
20	20	20	0.051	17	34	0.122	15	45	0.101	13	52	0.111
40	29	29	0.051	24	48	0.134	20	60	0.117	17	68	0.121
60	40	40	0.072	32	64	0.147	28	84	0.146	24	96	0.167
80	45	45	0.087	33	66	0.173	28	84	0.162	25	100	0.168
100	56	56	0.112	52	104	0.191	44	132	0.198	43	172	0.232
120	65	65	0.149	46	92	0.188	44	132	0.227	36	144	0.251
150	74	74	0.171	51	102	0.210	42	126	0.223	38	152	0.251
200	93	93	0.206	77	154	0.322	65	195	0.355	66	264	0.455
400	167	167	0.557	123	246	0.692	107	321	0.796	110	440	0.976
600	242	242	1.959	153	306	1.392	117	351	1.244	102	408	1.258
800	310	310	2.992	213	426	2.143	188	564	2.130	199	796	2.586
0 GMRES iterations 0 200 - 20		E(MG) P(1)+CS P(2)+CS P(3)+CS	SL(MG) SL(MG) SL(MG)		8	Lime (s)	3 2.5 2 - *- - *- - *- - *- - *- - *- - *- -	- CSL(N - FP(1) - FP(2) - FP(3)	AG) +CSL(MG +CSL(MG +CSL(MG	+ +) +)	0	
5 International Strategy 100 - 00 - 00 - 00 - 00 - 00 - 00 - 00		200	300 400	500	-+	- 0	1 - .5 -		300 4	■ * *	. 600	700 80

FIG. 4.4. Model problem 2: Number of GMRES iterations (left) and time elapsed (right) for computing the shifted Laplacian with a multigrid cycle.

Wavenumber

F(1,1)-cycle for the multigrid solve of the shifted Laplace preconditioner and shift parameter $\varepsilon = 0.5k^2$. The iteration is stopped when the residual is reduced by a factor of 10^{-8} . The results are shown in Figure 4.5 (top left). The Faber preconditioner does not reduce the number of matrix-vector products, but since there is only a slight increase in matrix-vector products, there may be a potential speedup from the reduction in the number of vector operations in GMRES. This will be explored in the subsequent experiments. We also see that there is no significant gain in using the Faber preconditioner of degree higher than 1. Repeating the experiment with full GMRES instead of GMRES(20) yields very similar results; see Figure 4.5 (top right). For k = 100, the results are similar (see Figure 4.5 (bottom)), and they show that no significant gain is obtained from a higher-degree truncated Faber series with only a slight increase in the number of matrix-vector products compared to the standalone CSL preconditioner.

Next, we solve model problem 3 for different wavenumbers. First, we use GMRES(20) and compare the standalone CSL preconditioner and the CSL combined with the Faber preconditioner of degrees 1 and 2. The number of GMRES(20) iterations and timings for increasing wavenumbers are presented in Figure 4.6 (top) and in Table 4.5. We can see that, with respect to the standalone CSL preconditioner, there is a decrease of 40% to 45% in the number of iterations ((iter. CSL - iter. CSL+FP(n))/iter. CSL) when applying the Faber

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FIG. 4.5. Model problem 3 with k = 50 (top) and k = 100 (bottom): Relative residual as a function of the number of matrix-vector products when preconditioning by the CSL and the Faber preconditioner with varying degrees. Left: GMRES(20). Right: full GMRES.

preconditioner of degree 1. With the Faber preconditioner of degree 2, the number of iterations is further reduced. However, there is no speedup, which is due to the fact that the number of matrix-vector products (multigrid solves of the CSL preconditioner) is not reduced. We repeat this experiment with full GMRES and display the results in Figure 4.6 (bottom) and Table 4.6. Similarly as before, applying the Faber preconditioner of degree 1 leads to a reduction in the number of iterations by 40% to 45% in comparison with the standalone CSL, and this translates to a speedup of approximately 33% to 35% for the larger wavenumbers.

4.4. Model problem 4. We now consider a Helmholtz problem in two dimensions with variable wavenumber and absorbing boundary conditions, similar to the ones proposed in [20, 38]:

$$\begin{aligned} -\Delta u - k^2 u &= f & \text{ in } \Omega = (0,1)^2, \\ \partial_n u - iku &= 0 & \text{ in } \partial \Omega, \end{aligned}$$

where f is a point source centered at (1/2, 1/2). The space-dependent wavenumber is

$$k(x,y) = \begin{cases} (4/3)k_{\text{ref}} & \text{if} \qquad 0 \le y < 0.2x + 0.2, \\ k_{\text{ref}} & \text{if} \qquad 0.2x + 0.2 \le y < -0.2x + 0.8, \\ 2k_{\text{ref}} & \text{if} -0.2x + 0.8 \le y < 1, \end{cases}$$



FIG. 4.6. Model problem 3 with GMRES(20) (top) and full GMRES (bottom): Number of iterations (left) and time in seconds (right) as a function of the wavenumber when preconditioning with the standalone CSL and the CSL combined with the Faber preconditioner with degrees 1 and 2.

where $k_{\text{ref}} \in \mathbb{R}$. This problem is discretized with second-order finite differences on a uniform grid with 12 points per wavelength [46].

Since restarted GMRES performs very poorly for this problem, we only use full GMRES and compare the standalone CSL preconditioner with the CSL combined with the Faber preconditioner of degrees 1 and 2. The number of GMRES iterations and timings for increasing wavenumbers are given in Figure 4.7 and in Table 4.7. As in the case of a constant wavenumber, we see that compared to the CSL preconditioner, applying the Faber preconditioner of degree 1 leads to a decrease in the number of iterations of approximately 40% and a reduction in CPU time of around 20% (and up to 40% for certain k).

5. Conclusions. In this paper we have introduced the Faber preconditioner for the discretized Helmholtz equation preconditioned with the complex shifted Laplacian. The Faber preconditioner is a polynomial preconditioner constructed by truncating the Faber series of 1/z on a non-convex 'bratwurst' set, which contains the eigenvalues of the CSL-preconditioned Helmholtz matrix. The Faber polynomials can be computed efficiently with a three-term recurrence, and we give analytic expressions for the coefficients of the Faber series. Finally, we show that the low-order Faber preconditioners are indeed nonsingular operators.

Our numerical experiments in 1D and 2D with constant and varying wavenumbers show that the Faber preconditioner typically reduces the number of GMRES iterations. When using

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TABLE 4.5

Model problem 3 with GMRES(20): Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with a multigrid cycle.

k		CSL(N	AG)	FP	(1)+CS	L(MG)	FP(2)+CSL(MG)			
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	
5	12	12	0.012	10	20	0.010	9	27	0.007	
20	24	24	0.037	14	28	0.037	11	33	0.043	
50	66	66	0.953	48	96	1.161	79	237	2.674	
80	86	86	1.219	46	92	1.117	43	129	1.482	
110	93	93	5.898	51	102	5.635	37	111	5.828	
140	140	140	8.939	86	172	9.398	68	204	10.660	

TABLE	4.6

Model problem 3 with full GMRES: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with a multigrid cycle.

k		CSL(N	4G)	FP	(1)+CS	L(MG)	FP(2)+CSL(MG)			
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	
5	12	12	0.013	10	20	0.011	9	27	0.007	
20	23	23	0.044	14	28	0.045	11	33	0.046	
50	57	57	1.298	38	76	1.125	37	111	1.443	
80	77	77	1.992	44	88	1.335	38	114	1.490	
110	90	90	10.739	50	100	7.135	36	108	6.481	
140	131	131	19.247	79	158	12.574	63	189	12.256	

full GMRES, we usually observe a reduction in the number of iterations and in CPU time, in particular for higher wavenumbers, which is due to the reduced number of vector operations.

As an outlook, the Faber preconditioner could be combined with deflation techniques. It is known [25] that the eigenvalues of the deflated CSL-preconditioned Helmholtz operator are shifted away from the origin and still lie in the disk $\{z \in \mathbb{C} : |z - 1/2| \le 1/2\}$, ensuring that the Faber preconditioner can be applied.

Appendix A. Boundary rotation.

In this section we discuss the boundary rotation of our inclusion set. We begin with the definition of the total rotation of a curve. The *total rotation* of a smooth Jordan curve Γ , parametrised by $\gamma : [a, b] \to \mathbb{C}$ with $\gamma'(t) \neq 0$, is

(A.1)
$$V(\Gamma) = \int_{a}^{b} \left| \frac{d}{dt} \arg(\gamma'(t)) \right| dt,$$

where $[a, b] \ni t \mapsto \arg(\gamma'(t))$ denotes a smooth branch of the argument. Geometrically, the total rotation is the total variation of the angle θ between the tangent to Γ and the positive real line and can also be written in the form $V = \int_{\Gamma} |d\theta|$. This definition can be extended to Jordan arcs by integrating along each "side" and "around the corners" (see [14]) and to arbitrary Jordan curves by a limiting process; see [32, 44]. For a compact set *E* bounded by a Jordan curve, we define $V = V(E) = V(\partial E)$, and *E* has *bounded boundary rotation* if $V < \infty$; see [14, 31, 32, 44]. We have $V \ge 2\pi$, and $V = 2\pi$ if and only if *E* is convex.

The set $\Omega = \Omega(\lambda, \phi, \sigma)$ in Theorem 3.2 has bounded boundary rotation; see [31, Lemma 3.5]. Then $E = \frac{1}{2}\Omega + \frac{1}{2}$ has the same boundary rotation, $V(E) = V(\Omega)$, which is easily seen from (A.1). Let us discuss $V(\Omega)$.



FIG. 4.7. Model problem 4: Number of GMRES iterations (left) and time in seconds (right) as a function of the wavenumber k_{ref} for the 2D wedge problem preconditioned by the standalone CSL and the CSL combined with the Faber preconditioner with degrees 1 and 2.

TABLE 4.7

Model problem 4: Number of GMRES iterations, matrix-vector products, and timings for computing the shifted Laplacian with a MG cycle.

$k_{\rm ref}$		CSL(N	4G)	FP	(1)+CS	L(MG)	FP(2)+CSL(MG)			
	Iter.	MV	Time(s)	Iter.	MV	Time(s)	Iter.	MV	Time(s)	
5	13	13	0.031	11	22	0.036	9	27	0.040	
10	21	21	0.106	14	28	0.108	10	30	0.107	
20	34	34	0.464	20	40	0.410	17	51	0.467	
30	55	55	3.426	31	62	2.673	24	72	2.823	
40	75	75	5.219	48	96	4.357	45	135	5.452	
50	84	84	6.269	52	104	4.767	45	135	5.476	
60	138	138	58.220	79	158	36.349	60	180	34.130	

For $\sigma = 1$, the set $\Omega = \Omega(\lambda, \phi, 1)$ is the circular arc $\{\lambda e^{i\beta} : \frac{\phi}{2} \leq \beta \leq 2\pi - \frac{\phi}{2}\}$ (see [31, Example 3.2]), which has interior angle $2\pi - \phi$. Its boundary rotation then is

$$V = 6\pi - 2\phi.$$

See also [6, p. 200] for the boundary rotation of an annular sector.

For $1 < \sigma < P$, the set Ω has an analytic boundary which can be parametrized by $\gamma(t) = \psi(e^{it}), 0 \le t \le 2\pi$. Then,

$$\frac{d}{dt}\arg(\gamma'(t)) = 1 + \frac{d}{dt}\arg(\psi'(e^{it})) = 1 + \operatorname{Re}\left(\frac{e^{it}\psi''(e^{it})}{\psi'(e^{it})}\right)$$

The last equality is well known and can be established as follows:

$$\frac{d}{dt} \arg(\psi'(e^{it})) = \frac{d}{dt} \operatorname{Re}(-i \log|\psi'(e^{it})| + \arg(\psi'(e^{it}))) = \frac{d}{dt} \operatorname{Re}(-i \log(\psi'(e^{it})))$$
$$= \operatorname{Re}\left(\frac{e^{it}\psi''(e^{it})}{\psi'(e^{it})}\right).$$

Hence, the boundary rotation of Ω is

(A.2)
$$V = \int_0^{2\pi} \left| 1 + \operatorname{Re}\left(\frac{e^{it}\psi''(e^{it})}{\psi'(e^{it})}\right) \right| dt.$$



FIG. A.1. Boundary rotation V/π for $\Omega(-1, \pi/2, \sigma)$ as a function of σ .

Using the map ψ from Theorem 3.2, we find

$$\frac{w\psi^{\prime\prime}(w)}{\psi^{\prime}(w)} = \frac{2w^2 + 2wS}{w^2 + w2S - \lambda^2 MN - \lambda(M+N)S} - \frac{2w}{w+S}$$

where $S = \lambda \frac{NM-1}{N-M}$. Together with (A.2), it is easy to compute the boundary rotation V numerically. Figure A.1 displays the computed values V/π for the sets $\Omega(-1, \pi/2, \sigma)$ with $1 < \sigma < P = 1.4966$ (rounded to four digits); see Figure 3.2 for the shape of Ω .

Appendix B. Proof of Proposition 3.4.

Proof of Proposition 3.4. Recall the expressions (3.7) for the Faber polynomials and (3.9) for the coefficients. The truncated Faber series of degree n = 1 is

$$s_1(z) = a_1 F_1(z) + a_0 F_0(z) = -\frac{\rho - M}{\rho^2} F_1(z) + \frac{\rho - M}{\rho} = -\frac{\rho - M}{\rho^2} (F_1(z) - \rho)$$

Thus, the unique zero z_1 of s_1 is the zero of

$$F_1(z) - \rho = 2(N - M)z - 2N - \frac{MN - 1}{N - M} - \rho,$$

i.e.,

$$z_1 = \frac{2N + \rho + \frac{MN-1}{N-M}}{2(N-M)}.$$

We have $z_1 > 1$ since the numerator satisfies

$$2N + \rho + \frac{MN - 1}{N - M} > 3N + \frac{MN - 1}{N - M} = 2N + \frac{N^2 - MN + MN - 1}{N - M}$$
$$> 2N \ge 2(N - M) > 0.$$

Thus, the zero of s_1 is not in the spectrum of AA_{ε}^{-1} and $s_1(AA_{\varepsilon}^{-1})$ is nonsingular. For n = 2, the truncated Faber series is

$$s_2(z) = a_2 F_2(z) + a_1 F_1(z) + a_0 F_0(z) = \frac{\rho - M}{\rho^3} (F_2(z) - \rho F_1(z) + \rho^2).$$

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Therefore, the zeros of s_2 are the zeros of $p(z) = F_2(z) - \rho F_1(z) + \rho^2$. We have

$$p(z) = az^2 + bz + c,$$

where the coefficients are given by

$$a = 4(N - M)^{2},$$

$$b = -8N^{2} + 4MN + 4 - 2\rho(N - M),$$

$$c = 4N^{2} - 2 - S^{2} + 2N\rho + \rho S + \rho^{2},$$

and where S = (MN - 1)/(N - M). Note that all coefficients are real.

We first show that p has no real roots and equivalently that the discriminant of p is negative. Let $\Delta = (b/2)^2 - ac$ be (one fourth) of the discriminant. Then

$$\Delta = (2 - 4N^2 + 2MN - \rho(N - M))^2 - 4(N - M)^2(4N^2 - 2 - S^2 + 2N\rho + \rho S + \rho^2).$$

Simplifying this expression (all computations in this paragraph have been performed with Mathematica 11.0) yields

$$\begin{split} \Delta &= 8 + 11 M^2 - 22 M N - 5 N^2 - 14 M^2 N^2 + 28 M N^3 - 6 N^4 \\ &- 6 N (M-N)^2 \sqrt{N^2 - 1}. \end{split}$$

The term with the square root is negative, so it is sufficient to show that the remaining term is also negative. The properties $0 \le M < 1 < N$ are not sufficient to prove that this first term is negative, so we work with the expressions of M and N in ϕ and σ from Theorem 3.2. Inserting these yields a positive factor times the term

$$-(1 - 8\sigma^{2} + \sigma^{4})\cos(\phi/2) + (1 + \sigma^{2})^{2}\cos(\phi) + 2(-7 + 3\sigma^{2} - 7\sigma^{4} + 8(\sigma^{4} - 1)\sin(\phi/4) + 2(\sigma^{4} - 1)\sin(3\phi/4)).$$

It is now sufficient to show that this term is negative for all $0 < \phi < 2\pi$ and admissible σ . For fixed $0 < \phi < 2\pi$, we substitute $\tau = \sigma^2$ to obtain a quadratic polynomial in τ :

$$q(\tau) = (-1 + 8\tau - \tau^2)\cos(\phi/2) + (1 + \tau)^2\cos(\phi) - 14 + 6\tau - 14\tau^2 + 16(\tau^2 - 1)\sin(\phi/4) + 4(\tau^2 - 1)\sin(3\phi/4).$$

The derivative of q is

$$q'(\tau) = (8 - 2\tau)\cos(\phi/2) + 2(1 + \tau)\cos(\phi) + 6 - 28\tau + 32\tau\sin(\phi/4) + 8\tau\sin(3\phi/4),$$

which has the unique zero

$$\tau_0 = \frac{4(\cos(\phi/8) + \sin(\phi/8))^4}{5 + 2\cos(\phi/2)}.$$

Then

$$q(\tau_0) = -\frac{6(7 + 4\cos(\phi/2))(\cos(\phi/8) + \sin(\phi/8))^4}{5 + 2\cos(\phi/2)} < 0$$

for all $0 < \phi < 2\pi$. The function q has a (global) maximum at this point since the coefficient of τ^2 in q is

$$-\cos(\phi/2) + \cos(\phi) - 14 + 16\sin(\phi/4) + 4\sin(3\phi/4)$$
$$= -2(5 + 2\cos(\phi/2))(\cos(\phi/8) - \sin(\phi/8))^4 < 0$$

for all $0 < \phi < 2\pi$. This shows that $q(\tau) < 0$ for all τ and ϕ , which implies that $\Delta < 0$ and thus that p has no real roots.

Hence the polynomial p has a pair of conjugate roots $\lambda, \overline{\lambda}$ and can be written in the form

$$p(z) = a\left(z^2 + \frac{b}{a}z + \frac{c}{a}\right) = a(z - \lambda)(z - \overline{\lambda}) = a(z^2 - 2\operatorname{Re}(\lambda)z + |\lambda|^2).$$

This gives the relations

$$\lambda \overline{\lambda} = |\lambda|^2 = \frac{c}{a}, \qquad \operatorname{Re}(\lambda) = -\frac{b}{2a}.$$

We show now that the roots satisfy $|\lambda - 1/2| > 1/2$. Since

$$\left|\lambda - \frac{1}{2}\right|^2 = \left(\operatorname{Re}(\lambda) - \frac{1}{2}\right)^2 + \operatorname{Im}(\lambda)^2 = |\lambda|^2 - \operatorname{Re}(\lambda) + \frac{1}{4} = \frac{c + \frac{b}{2}}{a} + \frac{1}{4}$$

and a > 0, we have that $|\lambda - 1/2| > 1/2$ is equivalent to $c + \frac{b}{2} > 0$. We compute

$$c + \frac{b}{2} = -S^2 + 2N\rho + \rho S + \rho^2 + 2MN - \rho(N - M)$$
$$= (\rho^2 - S^2) + \rho(S + N + M) + 2MN.$$

To prove that c+b/2>0, we show next that S+N+M and ρ^2-S^2 are positive. A short computation yields

$$S + N + M = \frac{N^2 - 1 + M(N - M)}{N - M} > 0$$

since $N > 1 > M \ge 0$. Moreover, we have

$$\rho^2 - S^2 = 2N\sqrt{N^2 - 1} + \frac{(2N^2 - 1)(N - M)^2 - (MN - 1)^2}{(N - M)^2}.$$

Expanding and rearranging the numerator of the last term, we obtain

$$\begin{aligned} (2N^2-1)(N^2-2MN+M^2) - (M^2N^2-2MN+1) \\ &= (N^2-1)(2(N-M)^2+(1-M^2)) > 0, \end{aligned}$$

since $0 \le M < 1 < N$. Therefore $\rho^2 - S^2 > 0$, and, with the previous calculation, this implies that c + b/2 > 0. We conclude that the roots of s_2 are outside the disk $|z - \frac{1}{2}| \le \frac{1}{2}$, so they are not eigenvalues of AA_{ε}^{-1} , and $s_2(AA_{\varepsilon}^{-1})$ is nonsingular. \Box

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