Preconditioners for ill-conditioned Toeplitz matrices *

Daniel Potts and Gabriele Steidl

\textsuperscript{1}Institut für Mathematik, Medizinische Universität zu Lübeck, Wallstr. 40, D-23560 Lübeck, Germany. email: potts@math.mv-luebeck.de
\textsuperscript{2}Fakultät für Mathematik und Informatik, Universität Mannheim, D-68131 Mannheim, Germany. email: steidl@kiwi.math.uni-mannheim.de

\textbf{Abstract.}

This paper is concerned with the solution of systems of linear equations $A_N x = b$, where $\{A_N\}_{N \in \mathbb{N}}$ denotes a sequence of positive definite Hermitian ill-conditioned Toeplitz matrices arising from a (real-valued) nonnegative generating function $f \in C_{2\pi}$ with zeros. We construct positive definite Hermitian preconditioners $M_N$ such that the eigenvalues of $M_N^{-1} A_N$ are clustered at 1 and the corresponding PCG-method requires only $O(N \log N)$ arithmetical operations to achieve a prescribed precision. We sketch how our preconditioning technique can be extended to symmetric Toeplitz systems, doubly symmetric block Toeplitz systems with Toeplitz blocks and non-Hermitian Toeplitz systems. Numerical tests confirm the theoretical expectations.

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\textbf{1 Introduction.}

Systems of linear equations $A_N x = b$

with positive definite Hermitian Toeplitz matrices $A_N$ arise in a variety of applications in mathematics and engineering (see [11] and the references therein). Along with stabilization techniques for direct fast and superfast Toeplitz solvers, preconditioned conjugate gradient methods (PCG-methods) and other iterative methods have attained much attention during the last years. As essential computational effort, the CG-method requires the multiplication of a vector with the matrix $A_N$ in each iteration step. For Toeplitz matrices $A_N$, the multiplication with a vector can be computed with $O(N \log N)$ arithmetical operations by fast Fourier transforms (FFT). The number of iteration steps of the CG-method depends on the distribution of the eigenvalue of $A_N$. In particular, it holds (see [1, p.573]

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THEOREM 1.1. Let $A_N$ be a positive definite Hermitian $(N, N)$-matrix which has $p$ and $q$ isolated large and small eigenvalues, respectively:

$$0 < \lambda_1 \leq \lambda_2 \leq \ldots \leq \lambda_q < \lambda_{q+1} \leq \ldots \leq \lambda_{N-p} \leq b$$

$$\lambda_{N-p+1} \leq \lambda_{N-p+2} \leq \ldots \leq \lambda_N \quad (0 < a < b < \infty).$$

Let $[x]$ denote the smallest integer $\geq x$. Then the CG-method for the solution of $A_N x = b$ requires at most

$$n = \left\lfloor \frac{\ln \frac{2}{r} + \sum_{k=1}^{q} \frac{b}{\lambda_k}}{\ln \left( \frac{1 + \left( \frac{a}{b} \right)^{1/2}}{1 - \left( \frac{a}{b} \right)^{1/2}} \right)} \right\rfloor + p + q$$

iteration steps to achieve precision $\tau$, i.e.

$$\frac{||x_n - x||_A}{||x_0 - x||_A} \leq \tau,$$

where $||x||_A := \sqrt{x^T A_N x}$ and where $x_n$ denotes the numerical solution after $n$ iteration steps.

In the sequel, we denote by $C_{2\pi}$ the space of $2\pi$-periodic, real-valued continuous functions.

Let $\{\sigma_k^N\}_{k=1}^N$ be a sequence of real numbers and let $\gamma_N(\epsilon)$ denote the number of those among $\sigma_k^N$ ($k = 1, \ldots, N$) which are outside the interval $(p - \epsilon, p + \epsilon)$. If $\gamma_N(\epsilon) < K(\epsilon)$, where $K(\epsilon)$ is independent of $N$, then we say that the values $\sigma_k^N$ are clustered at $p$ [32]. If the eigenvalues of a sequence of $(N, N)$-matrices $A_N$ are clustered at 1, then the CG-method converges superlinearly (see [13]).

For a sequence of $(N, N)$-Toeplitz matrices $A_N = A_N(f)$ ($N \in \mathbb{N}$) generated by a function $f \in C_{2\pi}$, it is well-known that the eigenvalues are distributed as $f$ [32, 17]. Let

$$f_{\min} := \min\{f(x) : x \in [0, 2\pi]\}, \quad f_{\max} := \max\{f(x) : x \in [0, 2\pi]\}.$$

Then the eigenvalues of $A_N(f)$ are contained in $[f_{\min}, f_{\max}]$. If $f > 0$, then by Theorem 1.1 the number of iteration steps of the CG-method to achieve a prescribed precision is independent of $N$ and the CG-method requires only $O(N \log N)$ arithmetical operations.

The situation changes completely, if we allow $f \geq 0$ to have zeros. In this case, the CG-method converges very slow with increasing $N$. To accelerate the convergence of the CG-method, several authors proposed preconditioners for Toeplitz systems. Clearly, the multiplication of any vector with the preconditioned matrix should also only require $O(N \log N)$ arithmetical operations. Therefore, two types of preconditioners were mainly exploited for linear Toeplitz systems, namely so-called “Strang-preconditioners” [13, 30, 14]

$$M_N(S_N f, F_N) := F_N \text{diag} \left( \left( \frac{2\pi j}{N} \right) \right)_{j=0}^{N-1} F_N,$$
where $S_N f$ denotes the $(N-1)$-th Fourier sum of $f$ and optimal preconditioners
\cite{15}
(1.3)
$$M_N^0(F_N) := F_N \delta(\tilde{F}_N A_N F_N) \tilde{F}_N,$$
where $\delta(A) := \text{diag}(a_{kk})_{k=0}^{N-1}$ and $a_{kk}$ are the diagonal entries of $A$. Here $F_N$
denotes the $N$-th Fourier matrix
$$F_N := \frac{1}{\sqrt{N}} (e^{-2\pi ijk/N})_{j,k=0}^{N-1}.$$

If $f > 0$, then both preconditioners $M_N$ are positive definite and the eigenvalues
of the preconditioned matrices $M_N^{-1} A_N$ are clustered at $1$. The same holds if we
replace $F_N$ by other unitary matrices, for example by the product of $F_N$ with a
unitary diagonal matrix which leads to $\omega$-circulant preconditioners \cite{10, 19} or, if
$A_N$ is real-valued, by unitary trigonometric matrices as the sine–I transform \cite{5}
which results in so-called $\tau$-preconditioners, the Hartley transform \cite{7} or other
trigonometric transforms \cite{8, 21}.

Unfortunately, if $f \geq 0$ has zeros, then the above preconditioners do not work
in general. The Strang–preconditioners are not positive definite for arbitrary
$f \in C_{2\pi}$. The convergence of the PCG-method with the above optimal
preconditioners is not independent of $N$ also in the $\tau$-case, if $f$ has zeros of higher
order \cite{6, 4}. An alternative choice are banded preconditioners belonging to the
Toeplitz or to the $\tau$-class which lead to satisfactory results \cite{5, 9, 10, 26, 4},
multigrid methods \cite{16} or "improved circulants" \cite{31}.

In this paper, we propose simple positive definite $\omega$–circulant preconditioners.
In particular, if $f(2\pi j/N) > 0$ for all $j = 0, \ldots, N-1$, then we obtain our
preconditioners by replacing $S_N f$ in (1.2) by $f$. In Section 3, we prove that
our preconditioners lead to superlinear convergence of the corresponding PCG-
method and that the number of PCG–iterations for reaching a fixed tolerance is
independent of $N$.

Our idea can be extended to (real) symmetric Toeplitz matrices, non-Hermitian
Toeplitz matrices and doubly symmetric block Toeplitz matrices with
Toeplitz blocks. We sketch various generalizations in Section 4. Writing
this paper, we became aware of the preprint \cite{20} of T. Huckle located at his home
page, where the author suggests a trigonometric preconditioner with respect to
the discrete sine transform of type I which is similar to our trigonometric
preconditioners in Section 4. However, our initial approach in the complex case
and our proofs are different from \cite{20}.

Numerical tests for Hermitian and symmetric Toeplitz matrices as well as for
non-symmetric Toeplitz matrices and doubly symmetric block Toeplitz matrices
with Toeplitz blocks in Section 5 demonstrate the quality of our new preconditioners.

2 Construction of preconditioners

Let $C_{2\pi}$ and $L^p_{2\pi}$ ($1 \leq p < \infty$) denote the Banach spaces of $2\pi$–periodic
continuous functions and of $2\pi$–periodic Lebesgue measurable functions with
finite integral \( \int_0^{2\pi} |f(x)|^p \, dx \), respectively. By \( o_N \), we denote the vector consisting of \( N \) zeros and by \( I_N \) the \((N,N)\)-identity matrix.

We are interested in the solution of Hermitian Toeplitz systems

\[
A_N(f) \mathbf{x} = b, \quad A_N(f) := (a_{j-k})_{j,k=0}^{N-1},
\]

where the sequence \( \{A_N(f)\}_{N=1}^\infty \) of Toeplitz matrices is generated by a nonnegative function \( f \in C_{2\pi} \), i.e.

\[
a_k = a_k(f) := \frac{1}{2\pi} \int_0^{2\pi} f(x) e^{-ikx} \, dx.
\]

Then we obtain for \( \mathbf{u} = (u_j)_{j=0}^{N-1} \in \mathbb{C}^N \) that

\[
\bar{\mathbf{u}}' A_N(f) \mathbf{u} = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} a_j u_k a_{j-k} = \frac{1}{2\pi} \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} a_j u_k \int_0^{2\pi} f(x) e^{-i(j-k)x} \, dx = \frac{1}{2\pi} \int_0^{2\pi} \left| \sum_{k=0}^{N-1} u_k e^{ikx} \right|^2 f(x) \, dx \geq 0
\]

such that the Toeplitz matrices \( A_N(f) \) are positive semidefinite. Moreover, if \( f > 0 \) on a set of positive Lebesgue measure, then following lemma states that the matrices \( A_N(f) \) are positive definite such that (2.1) can be solved by the CG-method.

**Lemma 2.1.** Let \( f \in L^2([0,2\pi]) : f(x) > 0 \) has a positive Lebesgue measure. Then the corresponding Toeplitz matrices \( A_N(f) \) are positive definite.

Lemma 2.1 was proved in [9]. However, the proof is very short such that we include it in this paper.

**Proof.** Let \( N \in \mathbb{N} \) be fixed. By the above considerations, it remains to show that \( 0 \) is not eigenvalue of \( A_N(f) \). Assume that \( A_N(f) \) has eigenvalue \( 0 \). Then there exists \( \mathbf{u} \in \mathbb{C}^N \) \( \mathbf{u} \neq o_N \) such that

\[
\bar{\mathbf{u}}' A_N(f) \mathbf{u} = \frac{1}{2\pi} \int_0^{2\pi} \left| \sum_{k=0}^{N-1} u_k e^{ikx} \right|^2 f(x) \, dx = 0.
\]

Since the integrand is nonnegative almost everywhere, the integrand must be zero almost everywhere. Consequently,

\[
\left| \sum_{k=0}^{N-1} u_k e^{ikx} \right| = 0
\]
on the set \( \{ x \in [0, 2\pi] : f(x) > 0 \} \) of positive Lebesgue measure. But this implies the contradiction \( \mathbf{u} = \mathbf{0}_N \). \( \square \)

By Theorem 1.1, the convergence of the CG–method depends on the distribution of the eigenvalues of \( A_N(f) \). Unfortunately, if the generating \( f \in C_{2\pi} \) has zeros, then the CG–method converges very slow. To accelerate the convergence of the CG–method we are looking for suitable preconditioners \( M_N(f) \) of \( A_N(f) \). Having Theorem 1.1 in mind, we want to find a Hermitian positive definite matrix \( M_N(f) \) such that the eigenvalues of \( M_N(f)^{-1} A_N(f) \) are bounded from below by a positive constant independent of \( N \) and the number of isolated eigenvalues of \( M_N(f)^{-1} A_N(f) \) is independent of \( N \).

For the construction of \( M_N(f) \) we consider (2.2). In the following, we assume that \( f \) has only a finite number of zeros. Then we can choose an equispaced grid

\[
x_l := \frac{2\pi l}{N} + w \quad (w \in \left[ 0, \frac{2\pi}{N} \right] ; \ l = 0, \ldots, N-1)
\]

such that

\[
f(x_l) > 0 \quad (l = 0, \ldots, N-1).
\]

Approximating the integral on the right–hand side of (2.2) by the trapezoidal rule with respect to the above grid, we obtain

\[
(2.4) \quad \bar{u}' A_N(f) \mathbf{u} = \frac{1}{2\pi} \int_0^{2\pi} N^{-1} \sum_{k=0}^{N-1} u_k e^{ikx} \int_0^{2\pi} f(x) \, dx
\]

\[
\approx \frac{1}{N} \sum_{l=0}^{N-1} \left| \sum_{k=0}^{N-1} u_k e^{ikx_l} \right|^2 f(x_l)
\]

\[
= \sum_{l=0}^{N-1} f(x_l) \frac{1}{\sqrt{N}} \left( \sum_{k=0}^{N-1} \bar{u}_k e^{-2\pi ikj/N} e^{-ijw} \right) \times
\]

\[
\frac{1}{\sqrt{N}} \left( \sum_{k=0}^{N-1} u_k e^{2\pi ikl/N} e^{ikw} \right)
\]

\[
= (F_N \bar{W}_N \bar{u})' D_N \bar{F}_N \bar{W}_N \mathbf{u}
\]

\[
= \bar{u}' M_N(f) \mathbf{u}
\]

with the diagonal matrices

\[
W_N := \text{diag} \left( e^{-ikw} \right)_{k=0}^{N-1}, \quad D_N := \text{diag} \left( f(x_l) \right)_{l=0}^{N-1}
\]

and with

\[
(2.5) \quad M_N(f) = M_N(f, F_N) := W_N F_N D_N \bar{F}_N \bar{W}_N.
\]

By (2.3), the matrix \( M_N(f) \) is Hermitian and positive definite. A matrix of the form (2.5) is called an \( \alpha \)–circulant matrix and in the special case that \( w = \pi/N \) a skew–circulant matrix. Setting \( v := M_N(f)^{1/2} \mathbf{u} \), we get

\[
\bar{v}' M_N(f)^{-1/2} A_N(f) M_N(f)^{-1/2} v \approx \bar{v}' v
\]
such that by properties of the Rayleigh quotient, \( M_N(f) \) seems to be a good preconditioner of \( A_N(f) \). Indeed, using FFT, the multiplication with
\[
M_N(f)^{-1} = W_N F_N D_N^{-1} \tilde{F}_N \tilde{W}_N
\]
takes only \( O(N \log N) \) arithmetical operations. In the next section, we prove that the eigenvalues of \( M_N(f)^{-1} A_N(f) \) are clustered at 1.

We mention that our preconditioner \( M_N(f) \) is closely related to the Strang-preconditioner \( M_N(S_N f) = M_N(S_N f, F_N) \) in (1.2). By orthogonality of the functions \( e^{ijx} \) (\( j \in \mathbb{Z} \)) in \( L^2_2 \), it is easy to check that (2.2) can be replaced by
\[
a' A_N(f) u = \frac{1}{2\pi} \int_0^{2\pi} \sum_{k=0}^{N-1} u_k e^{ikx}^2 (S_N f)(x) \, dx
\]
with
\[
(S_N f)(x) := \sum_{j=-(N-1)}^{N-1} a_j e^{ijx}.
\]
Now the above quadrature formula (2.4) with \( w = 0 \) and with \( S_N f \) instead of \( f \) leads to the Strang-preconditioner. Clearly, if \( f \) is a trigonometric polynomial of degree \( < N \) and if \( f(2\pi l/N) > 0 \) \( (l = 0, \ldots, N-1) \), then \( M_N(S_N f) = M_N(f) \).

However, for arbitrary nonnegative functions \( f \in C_2 \), the matrix \( M_N(S_N f) \) may be not positive definite. This is one reason for the introduction of \( M_N(f) \).

3 Clustering of the eigenvalues of \( M_N(f)^{-1} A_N(f) \)

We rewrite (2.4) as
\[
\tilde{a}' A_N(f) \tilde{u} = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} \tilde{a}_j u_k a_{j-k}
\approx \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} \tilde{a}_j u_k \tilde{a}_{j-k} = \tilde{a}' M_N(f) \tilde{u}
\]
with
\[
\tilde{a}_k = \tilde{a}_k(f) := \frac{1}{N} \sum_{l=0}^{N-1} f(x_l) e^{-2\pi ilk/N} e^{-ikw}
\]
and ask for the approximation error. Assume that \( f \in C_2 \) is a function of bounded variation. Replacing \( f(x_l) \) by the Fourier series of \( f \) at \( x_l \), we obtain
\[
\tilde{a}_k = \frac{1}{N} \sum_{l=0}^{N-1} \sum_{j \in \mathbb{Z}} a_j e^{ijx_l} e^{-2\pi ilk/N} e^{-ikw}
= \sum_{j=0}^{N-1} a_j e^{-iwk} e^{ij} \left( \frac{1}{N} \sum_{l=0}^{N-1} e^{-2\pi ilk/N} e^{2\pi ij/N} \right)
\]
\[ + \sum_{j=0}^{N-1} \sum_{r \in \mathbb{Z} \setminus \{0\}} a_{j+rN} e^{-i\omega_k} e^{i\omega(j+rN)} \left( \frac{1}{N} \sum_{l=0}^{N-1} e^{-2\pi i l k/N} e^{2\pi i j/l} \right) \]
\[ = a_k + \sum_{r \in \mathbb{Z} \setminus \{0\}} a_{k+rN} e^{i\omega r N} . \]

This is the well-known aliasing effect. Set
\[ (3.3) \quad b_k = b_k(f) := \sum_{r \in \mathbb{Z} \setminus \{0\}} a_{k+rN}(f) e^{i\omega r N}. \]

Then it follows by (3.2) that
\[ (3.4) \quad A_N(f) = M_N(f) + B_N(f), \quad B_N(f) := -(b_{j-k})_{j,k=0}^{N-1}. \]

Thus
\[ (3.5) \quad M_N(f)^{-1} A_N(f) = I_N + M_N(f)^{-1} B_N(f). \]

Note that
\[ b_k(S_N f) = \begin{cases} a_{k-N}(f) & k = 1, \ldots, N - 1, \\ a_{k+N}(f) & k = -1, \ldots, 1 - N, \\ 0 & k = 0, \end{cases} \]

which describes the approximation error in case of the Strang–preconditioner.

**Lemma 3.1.** Let \( p_s \) be a nonnegative, real-valued trigonometric polynomial of degree \( \leq s \), where \( 2s \leq N \). Then at most \( 2s \) eigenvalues of \( M_N(p_s)^{-1} A_N(p_s) \) differ from 1.

**Proof.** By (3.3), it follows that \( b_k = 0 \) for \( |k| \leq N - 1 - s \). Consequently, \( B_N(f) \) has rank \( 2s \). Now the assertion follows by (3.5). \( \square \)

For the proof of our main theorem we need the following

**Lemma 3.2.** Let \( g \in C_{2\pi} \) be a nonnegative function, where the set \( \{ x \in [0, 2\pi] : g(x) > 0 \} \) has a positive Lebesgue measure. Furthermore, let \( h \in C_{2\pi} \) be a positive function with \( h_{\text{min}} > 0 \) and let \( f := g h \). Then, for any \( N \in \mathbb{N} \), the eigenvalues of \( A_N(g)^{-1} A_N(f) \) lie in the interval \([h_{\text{min}}, h_{\text{max}}]\).

Lemma 3.2 was proved for example in [5]. For a more sophisticated version see [28, 27]. We want to give the following very simple proof.

**Proof.** Applying the theorem of mean in
\[ \tilde{u}' A_N(f) u = \frac{1}{2\pi} \int_0^{2\pi} \left| \sum_{k=0}^{N-1} u_k e^{ikx} \right|^2 f(x) \, dx \]
we obtain that
\[ \tilde{u}' A_N(f) u = \tilde{h}_s \frac{1}{2\pi} \int_0^{2\pi} \left| \sum_{k=0}^{N-1} u_k e^{ikx} \right|^2 g(x) \, dx \]
with \( h_* \in [h_{\min}, h_{\max}] \). This can be rewritten as

\[
\bar{u}' A_N(f) \, u = h_* \bar{u}' A_N(g) \, u.
\]

By Lemma 2.1, the matrix \( A_N(g) \) is positive definite such that for \( u \neq o_N \)

\[
h_* = \frac{\bar{u}' A_N(f) \, u}{\bar{u}' A_N(g) \, u}.
\]

By properties of the Rayleigh quotient this yields the assertion. \( \Box \)

In the following, we restrict our attention to nonnegative functions \( f \in C_{2\pi} \) having a zero of even order \( 2s \) \((s \in \mathbb{N})\) in \( x = 0 \). The clustering of the eigenvalues of \( M_N(f)^{-1} A_N(f) \) for arbitrary functions

\[
f(x) = (x - x_1)^{2s_1} \ldots (x - x_m)^{2s_m} \tilde{f}(x), \quad (\tilde{f} > 0)
\]

follows in a similar way.

With \( f \in C_{2\pi} \) or better with the order \( 2s \) of the zero \( x = 0 \) of \( f \), we associate the nonnegative trigonometric polynomial

\[
p_s(x) := (2 - 2 \cos x)^s = (2 - e^{ix} - e^{-ix})^s
\]

of degree \( s \) which has a zero of the same order \( 2s \) in \( x = 0 \).

Now we can prove our main result.

**THEOREM 3.3.** Let \( f \in C_{2\pi} \) be a nonnegative function with a zero of order \( 2s \) \((s \in \mathbb{N})\) in \( x = 0 \). Let \( A_N(f) \) denote the corresponding Toeplitz matrices with preconditioners \( M_N(f) \) defined by (2.5). Then the matrices \( M_N(f)^{-1} A_N(f) \) have the following properties:

i) The eigenvalues of \( M_N(f)^{-1} A_N(f) \) are bounded from below by a positive constant independent of \( N \).

ii) Let \( p_s \) denote the associated trigonometric polynomial (3.6) of \( f \) and let \( h := f/p_s \). Then, for \( N \geq 2s \), at most \( 2s \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are not contained in the interval \([h_{\min}, h_{\max}]\).

iii) The eigenvalues of \( M_N(f)^{-1} A_N(f) \) are clustered at \( 1 \).

**PROOF.** In this proof, we denote by \( R_N(m) \) arbitrary \((N,N)\)-matrices of rank \( m \).

1. To prove ii), we use the decomposition

\[
\frac{\bar{u}' A_N(f) \, u}{\bar{u}' M_N(f) \, u} = \frac{\bar{u}' A_N(f) \, u}{\bar{u}' A_N(p_s) \, u} \frac{\bar{u}' A_N(p_s) \, u}{\bar{u}' M_N(f) \, u} \quad (u \neq o_N).
\]

By Lemma 3.2 and since \( A_N(p_s) \) and \( M_N(f) \) are positive definite, it follows that

\[
h_{\min} \frac{\bar{u}' A_N(p_s) \, u}{\bar{u}' M_N(f) \, u} \leq \frac{\bar{u}' A_N(f) \, u}{\bar{u}' M_N(f) \, u} \leq h_{\max} \frac{\bar{u}' A_N(p_s) \, u}{\bar{u}' M_N(f) \, u}.
\]
By (3.4) and Lemma 3.1, we conclude that

\[ A_N(p_s) = M_N(p_s) + R_N(2s) \]

and consequently

\[ \frac{\vec{u}' A_N(f) u}{\vec{u}' M_N(f) u} \leq h_{\text{max}} \frac{\vec{u}' M_N(p_s) u}{\vec{u}' M_N(f) u} + h_{\text{max}} \frac{\vec{u}' R_N(2s) u}{\vec{u}' M_N(f) u}. \]

By construction of \( M_N(f) \) this can be rewritten as

\[ \frac{\vec{u}' (A_N(f) - h_{\text{max}} R_N(2s)) u}{\vec{u}' M_N(f) u} \leq h_{\text{max}} h_{\text{min}} (u \neq o_N). \]

Assume that \( R_N(2s) \) has \( s_1 \) positive eigenvalues. Then, by properties of the Rayleigh quotient and by Weyl's theorem [18, p.184] at most \( s_1 \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are larger than \( \frac{h_{\text{max}}}{h_{\text{min}}} \). Similarly, we obtain that by consideration of the left-hand side of (3.7) that at most \( 2s - s_1 \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are smaller than \( \frac{h_{\text{min}}}{h_{\text{max}}} \). Thus, at most \( 2s \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are not contained in \([h_{\text{min}}, h_{\text{max}}]\).

2. To prove i), we use the decomposition (see [4, 6])

\[ \frac{\vec{u}' A_N(f) u}{\vec{u}' M_N(f) u} = \frac{\vec{u}' A_N(f) u}{\vec{u}' A_N(p_s) u} \frac{\vec{u}' M_N(p_s) u}{\vec{u}' M_N(f) u} \frac{\vec{u}' A_N(p_s) u}{\vec{u}' M_N(p_s) u} (u \neq o_N). \]

By Lemma 3.2 and construction of \( M_N(f) \) we see that

\[ \frac{\vec{u}' A_N(f) u}{\vec{u}' M_N(f) u} \geq \frac{h_{\text{min}}}{h_{\text{max}}} \frac{\vec{u}' A_N(p_s) u}{\vec{u}' M_N(p_s) u}. \]

Since \( A_N(p_s) \) and \( M_N(p_s) \) are positive definite, it remains to show that there exists \( c < \infty \) such that

\[ \frac{\vec{u}' A_N(p_s) u}{\vec{u}' M_N(p_s) u} \geq \frac{1}{c} > 0. \]

By (3.4) this can be rewritten as

\[ \frac{\vec{u}' (A_N(p_s) - B_N(p_s)) u}{\vec{u}' A_N(p_s) u} \leq c \]

\[ 1 + \frac{\vec{u}' (-B_N(p_s)) u}{\vec{u}' A_N(p_s) u} \leq c. \]

The rest of the proof follows the same lines as the proof of Theorem 4.3 in [4], which was formulated for so-called \( \tau \)-preconditioners.
3. By definition, $h = f / p_s$ is a continuous positive function. Since the trigonometric polynomials are dense in $C_{2\pi}$, for all $\varepsilon > 0$, there exist a positive trigonometric polynomial $q$ of degree $n = n(\varepsilon)$ such that
\begin{equation}
q(x) - \frac{1}{2} \varepsilon h_{\min} \leq h(x) \leq q(x) + \frac{1}{2} \varepsilon h_{\min}
\end{equation}
for all $x \in [0, 2\pi)$. Thus, since $p_s \geq 0$,
\begin{equation}
q p_s - \frac{1}{2} \varepsilon h_{\min} p_s \leq f \leq q p_s + \frac{1}{2} \varepsilon h_{\min} p_s.
\end{equation}
Regarding (2.2), we obtain by the inequality of the right-hand side
\[ \tilde{u}^t A_N(f) u \leq \tilde{u}^t A_N(q p_s) u + \frac{1}{2} \varepsilon h_{\min} \tilde{u}^t A_N(p_s) u, \]
and further, since $M_N(f)$ is positive definite, for all $u \in \mathbb{C}^N$ ($u \neq \alpha_N$)
\begin{equation}
\frac{u^t A_N(f) u}{u^t M_N(f) u} \leq \frac{u^t A_N(q p_s) u}{u^t M_N(f) u} + \frac{1}{2} \varepsilon h_{\min} \frac{u^t A_N(p_s) u}{u^t M_N(f) u}.
\end{equation}
Now it holds by (3.4) and Lemma 3.1 that
\begin{equation}
A_N(p_s) = M_N(p_s) + R_N(2s).
\end{equation}
Moreover, we have by [3] that
\[ A_N(q p_s) = A_N(q) A_N(p_s) + R_N(2n + 2s). \]
By (3.11), this can be written as
\begin{equation}
A_N(q p_s) = (M_N(q) + R_N(2n)) (M_N(p_s) + R_N(2s)) + R_N(2n + 2s)
\end{equation}
with a Hermitian matrix $R_N(m)$ of rank $m \leq 4n + 4s + \min\{2n, 2s\}$. Substituting (3.11) and (3.12) in (3.10), we obtain
\[ \frac{\tilde{u}^t A_N(f) u}{\tilde{u}^t M_N(f) u} \leq \frac{\tilde{u}^t M_N(q) M_N(p_s) u}{\tilde{u}^t M_N(f) u} + \frac{\tilde{u}^t R_N(m) u}{\tilde{u}^t M_N(f) u} + \frac{1}{2} \varepsilon h_{\min} \frac{\tilde{u}^t M_N(p_s) u}{\tilde{u}^t M_N(f) u} \]
and since
\[ \frac{\tilde{u}^t M_N(p_s) u}{\tilde{u}^t M_N(f) u} \leq \frac{1}{h_{\min}} \]
further
\[ \frac{\tilde{u}^t [A_N(f) - R_N(\tilde{m})] u}{\tilde{u}^t M_N(f) u} \leq \frac{\tilde{u}^t M_N(q) M_N(p_s) u}{\tilde{u}^t M_N(f) u} + \frac{1}{2} \varepsilon. \]
with \( \bar{m} \leq m + 2s \). Setting \( v := M_N(p_{\bar{m}})^{1/2}u \) and using that \( M_N(f) = M_N(h) M_N(p_{\bar{m}}) \), we get

\[
\frac{\bar{u}' [A_N(f) - R_N(\bar{m})] u}{\bar{u}' M_N(f) u} \leq \frac{\bar{v}' M_N(q) v}{\bar{v}' M_N(h) v} + \frac{1}{2} \varepsilon.
\]

(3.13)

Finally, we have by (3.8) and by definition of \( M_N \), for all \( v \in \mathbb{C}^N \) (\( v \neq o_N \)) that

\[
\bar{v}' M_N(q) v \leq \bar{v}' M_N(h) v + \frac{1}{2} \varepsilon h_{\min} \bar{v}' v
\]

and further since \( 0 < \frac{\bar{v}' v}{\bar{v}' M_N(h) v} \leq \frac{1}{h_{\min}} \) that

\[
\frac{\bar{v}' M_N(q) v}{\bar{v}' M_N(h) v} \leq 1 + \frac{1}{2} \varepsilon.
\]

Using the above inequality in (3.13), we obtain

\[
\frac{\bar{u}' [A_N(f) - R_N(\bar{m})] u}{\bar{u}' M_N(f) u} \leq 1 + \varepsilon.
\]

Similarly, we conclude from the left-hand inequality of (3.9) that

\[
\frac{\bar{u}' [A_N(f) - R_N(\bar{m})] u}{\bar{u}' M_N(f) u} \geq 1 - \varepsilon.
\]

Consequently, at most \( \bar{m} \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are not contained in \([1 - \varepsilon, 1 + \varepsilon]\). This completes the proof. \( \Box \)

By Theorem 3.3, Theorem 1.1 and construction of \( M_N(f) \), our PCG-method converges superlinearly and requires only \( O(N \log N) \) arithmetical operations to achieve a prescribed precision.

**Remark 3.1.** Unfortunately, we cannot find a similar proof for nonnegative functions \( f \in C_{2\pi} \) having not only zeros of even order. The reason therefore is that there does not exist a nonnegative trigonometric polynomial which has a zero of odd order in \( x = 0 \). Consequently, we cannot produce an equivalent of (3.6). Our numerical tests show that our preconditioners work well also in the odd case. However, for the matrices \( A_N(f) \) generated by the function

\[
f(x) = \sqrt{2 - 2 \cos x} = |2 \sin \frac{x}{2}|,
\]

the number \( n \) of eigenvalues of \( M_N^{-1}(f) A_N(f) \) which are not contained in the interval \((1 - \varepsilon, 1 + \varepsilon)\) grows as follows:

\[
\begin{array}{c|c|c|c|c|c}
N & 32 & 64 & 128 & 256 & 512 \\
\hline
\varepsilon = 10^{-3} & 7 & 8 & 9 & 10 & 11 \\
\hline
\varepsilon = 10^{-5} & 10 & 12 & 13 & 15 & 17
\end{array}
\]

At first glance it seems that the eigenvalues of \( M_N(f)^{-1} A_N(f) \) are not clustered at 1.
4 Generalizations of the preconditioning technique

In this section, we sketch how our preconditioners can be generalized to the following settings:
- \( A_N(f) \) are (real) symmetric Toeplitz matrices,
- \( A_N(f) \) are non-Hermitian Toeplitz matrices,
- \( A_{M,N}(f) \) are doubly symmetric block Toeplitz matrices with Toeplitz blocks.

4.1 Preconditioners for symmetric Toeplitz matrices

First, we suppose in addition to Section 2 that the generating function \( f \in C_{2\pi} \) of the matrices \( A_N(f) \) is even. Then

\[
a_k = a_k(f) = \frac{2}{\pi} \int_0^\pi f(x) \cos kx \, dx
\]

and the Toeplitz matrices \( A_N(f) \in \mathbb{R}^{N,N} \) are symmetric. In this case, the multiplication of a vector with \( A_N(f) \) can be realized using fast trigonometric transforms instead of fast Fourier transforms. See Remark 4.1. In this way, complex arithmetic can be completely avoided in the iterative solution of (2.1). This is one of the reasons to look for preconditioners of type (2.5), where the Fourier matrix \( F_N \) is replaced by trigonometric matrices corresponding to fast trigonometric transforms.

In practice, four discrete sine transforms (DST) and four discrete cosine transforms (DCT) were applied (see [33]). Any of these eight trigonometric transforms can be realized with \( \mathcal{O}(N \log N) \) arithmetical operations (see for example [2], [29]). Likewise, we can define preconditioners with respect to any of these transforms. Here we refer to the extensive examinations in [23]. For the case \( f \geq 0 \), interesting results concerning \( r \)-preconditioners which are related to the sine-I transform are contained in [6, 4]. In particular, Di Benedetto suggested a banded preconditioner of the form \( M_N(p_r) \) with respect to the sine-I transform.

In this paper, we restrict our attention to the so-called DST-II and DCT-II, which are determined by the following transform matrices:

\[
\text{DCT-II : } C^{II}_N := \left( \frac{2}{N} \right)^{1/2} \left( \varepsilon^{N}_j \cos \left( \frac{j(k+1)\pi}{2N} \right) \right)_{j,k=0}^{N-1} \in \mathbb{R}^{N,N},
\]

\[
\text{DST-II : } S^{II}_N := \left( \frac{2}{N} \right)^{1/2} \left( \varepsilon^{N}_{j+1} \sin \left( \frac{(j+1)(k+1)\pi}{2N} \right) \right)_{j,k=0}^{N-1} \in \mathbb{R}^{N,N},
\]

where \( \varepsilon^{N}_j := 2^{-1/2} (k = 0, N) \) and \( \varepsilon^{N}_{j+1} := 1 (k = 1, \ldots, N-1) \). Moreover, we use the DCT-I with transform matrix

\[
\tilde{C}^{I}_{N+1} := \left( (\varepsilon^{N}_k)^2 \cos \frac{jk\pi}{N} \right)_{j,k=0}^{N}.
\]
The matrices \( C^I_N \) and \( S^I_N \) are orthogonal and \( \tilde{C}^I_{N+1} \) satisfies
\[
(4.1) \quad \tilde{C}^I_{N+1} \tilde{C}^I_{N+1} = \frac{N}{2} I_{N+1}.
\]

The eight trigonometric transforms are closely related to Toeplitz matrices [24]. In particular, it holds for the DCT–II and the DST–II:

**Lemma 4.1.** Let \( \text{stoop} a' \) and \( \text{shank} a' \) denote a symmetric Toeplitz matrix and a persymmetric Hankel matrix with first row \( a' \), respectively. Then there exist the following relations between trigonometric transforms and symmetric Toeplitz matrices:
\[
\begin{align*}
\left(C^I_N\right)^t D_1 C^I_N &= \frac{1}{2} \text{stoop}(a_0, \ldots, a_{N-1}) + \frac{1}{2} \text{shank}(a_1, \ldots, a_{N-1}, 0), \\
\left(S^I_N\right)^t D_2 S^I_N &= \frac{1}{2} \text{stope}(a_0, \ldots, a_{N-1}) - \frac{1}{2} \text{shank}(a_1, \ldots, a_{N-1}, 0)
\end{align*}
\]
with
\[
\begin{align*}
D_1 &:= \text{diag}(d_0, \ldots, d_{N-1})', \\
D_2 &:= \text{diag}(d_1, \ldots, d_N)', \\
d &= (d_0, \ldots, d_N)' := \tilde{C}^I_{N+1} (a_0, \ldots, a_{N-1}, 0)'.
\end{align*}
\]

For the proof see [24].

**Remark 4.1.** By Lemma 4.1, it follows that
\[
\text{stope}(a_0, \ldots, a_{N-1}) = \left(C^I_N\right)^t D_1 C^I_N + \left(S^I_N\right)^t D_2 S^I_N.
\]

Thus, if the vector \( d \) is precomputed by the DCT–I, then the multiplication of a vector with a symmetric Toeplitz matrix of size \( (N, N) \) requires two DCT–II, two DST–II and \( 2N \) real multiplications and can therefore be realized with \( O(N \log N) \) arithmetical operations (see also [24]).

Since for even \( f \in C_{2\pi} \), the \((N - 1)\)–th Fourier sum can be written as
\[
(S_N f)(z) = 2 \sum_{k=0}^{N-1} (\xi_k^N)^2 a_k \cos(kz),
\]
we obtain by Lemma 4.1 that
\[
\begin{align*}
A_N(f) &= \left(C^I_N\right)^t (2D) C^I_N - \text{shank}(a_1, \ldots, a_{N-1}, 0) \\
&= \left(C^I_N\right)^t \text{diag} \left((S_N f)(j\pi N)^{-1}\right)_{j=0}^{N-1} C^I_N - \text{shank}(a_1, \ldots, a_{N-1}, 0), \\
A_N(f) &= \left(S^I_N\right)^t (2D) S^I_N + \text{shank}(a_1, \ldots, a_{N-1}, 0) \\
&= \left(S^I_N\right)^t \text{diag} \left((S_N f)(j\pi N)^{-1}\right)_{j=1}^{N} S^I_N + \text{shank}(a_1, \ldots, a_{N-1}, 0).
\end{align*}
\]
Consequently, we introduce the Strang-type-preconditioners by [25]:

\[
\begin{align*}
\text{DCT – II}: & \quad M_N(S_N f, C_{N}^{II}) := (C_N^{II})' \text{diag} \left( S_N f \left( \frac{j \pi}{N} \right) \right)_{j=0}^{N-1} C_N^{II}, \\
\text{DST – II}: & \quad M_N(S_N f, S_{N}^{II}) := (S_N^{II})' \text{diag} \left( S_N f \left( \frac{j \pi}{N} \right) \right)_{j=1}^{N} S_N^{II}.
\end{align*}
\]

(4.4)

See also [21]. Again, if \( f \) has zeros, then it can not be assured that the Strang-type-preconditioners are positive definite. Therefore, we define similar to (2.5) the preconditioners

\[
\begin{align*}
\text{DCT – II}: & \quad M_N(f, C_{N}^{II}) := (C_N^{II})' \text{diag} \left( f \left( \frac{j \pi}{N} \right) \right)_{j=0}^{N-1} C_N^{II}, \\
\text{DST – II}: & \quad M_N(f, S_{N}^{II}) := (S_N^{II})' \text{diag} \left( f \left( \frac{j \pi}{N} \right) \right)_{j=1}^{N} S_N^{II}.
\end{align*}
\]

(4.5)

If \( f(j \pi/N) > 0 \) for all \( j = 0, \ldots, N - 1 \), then \( M_N(f, C_{N}^{II}) \) is positive definite.

If \( f(j \pi/N) > 0 \) for all \( j = 1, \ldots, N \), then \( M_N(f, S_{N}^{II}) \) is positive definite.

Note that independent of our results, T. Huckle [20] suggested a preconditioner of type (4.5) with respect to the DST–I.

Clearly, if \( f \) is a trigonometric polynomial of degree < \( N \), then the Strang-type-preconditioners (4.4) coincide with our preconditioners (4.5). Moreover, we have by (4.2) and (4.3) for trigonometric polynomials \( f = p \) of degree ≤ \( s \) (2\( s \) ≤ \( N \)) that

\[
A_N(p) = M_N(p, C_{N}^{II}) - R_N(2s) = M_N(p, S_{N}^{II}) + R_N(2s).
\]

Thus, we can prove in a completely similar way as in Section 3 the following

**Theorem 4.2.** Let \( f \in C_{2\pi} \) be an even nonnegative function with a zero of order 2\( s \) (\( s \in \mathbb{N} \)) at \( x = 0 \). Let \( A_N(f) \) denote the corresponding Toeplitz matrices with preconditioners \( M_N(f) = M_N(f, S_{N}^{II}) \) defined by (4.4). Then the matrices \( M_N(f)^{-1} A_N(f) \) have the following properties:

i) The eigenvalues of \( M_N(f)^{-1} A_N(f) \) are clustered at 1.

ii) Let \( p_x \) denote the associated trigonometric polynomial (3.6) of \( f \) and let \( h := f/p_x \). Then, for \( N \geq 2s \), at most 2\( s \) eigenvalues of \( M_N(f)^{-1} A_N(f) \) are not contained in the interval \( [\frac{h_{\min}}{h_{\max}}, \frac{h_{\max}}{h_{\min}}] \).

The PCG-method with our preconditioners can be realized in a more efficient way than the PCG-method with banded Toeplitz matrices as preconditioners [22, 26]:

**Remark 4.2.** Our PCG-method requires only two DCT–II, two DST–II and \( O(N) \) real multiplications in each iteration step. This can be seen for the preconditioner \( M_N(f, C_{N}^{II}) \) as follows: Instead of

\[
(C_N^{II})' E^{-1} C_N^{II} \left( (C_N^{II})' D C_N^{II} + (S_N^{II})' \tilde{D} S_N^{II} \right) x = (C_N^{II})' E^{-1} C_N^{II} b
\]
with $E := \text{diag} \left( f \left( \frac{j\pi}{N} \right) \right)_{j=0}^{N-1}$, we solve
\[
E^{-1} \left( D + C_{N}^{(1)} (S_{N}^{(1)})^t \tilde{D} S_{N}^{(1)} (C_{N}^{(1)})^t \right) \tilde{x} = \tilde{b}
\]
with $\tilde{x} := C_{N}^{(1)} x$ and $\tilde{b} := E^{-1} C_{N}^{(1)} b$. The vectors $d$, $b$ and $x$ can be precomputed and postcomputed, respectively. See also [19, 20].

4.2 Preconditioners for non-Hermitian Toeplitz matrices

Next, we are interested in the solution of systems of linear equations $A_{N}(f)x = b$ with regular, but non-Hermitian Toeplitz matrices $A_{N}(f)$. We intend to solve the normal equation
\[
\tilde{A}_{N}(f)A_{N}(f)x = \tilde{A}_{N}(f)b
\]
using the PCG-method. By [3], it holds that
\[
\tilde{A}_{N}(f)A_{N}(f) = A_{N}(|f|^2) + R_{N} + U_{N},
\]
with a low rank matrix $R_{N}$ and a matrix $U_{N}$ of small spectral norm. If $f = p$ is a trigonometric polynomial of degree $s$ ($2s \leq N$), then
\[
\tilde{A}_{N}(f)A_{N}(f) = A_{N}(|f|^2) + R_{N}(2s).
\]
Assume that $|f| \in C_{2\pi}$ has only a finite number of zeros. If $A_{N}(|f|^2)$ is Hermitian and if $|f(\frac{2\pi j}{N} + w)| > 0$ for a suitable $w \in [0, 2\pi/N)$ for all $j = 0, \ldots, N-1$, then we define our preconditioners by
\[
M_{N}(|f|^2, F_{N}) := W_{N} F_{N} \text{ diag} \left( |f(\frac{2\pi j}{N} + w)|^2 \right)_{j=0}^{N-1} W_{N}.
\]
If $A_{N}(|f|^2)$ is symmetric and if $|f(\frac{2\pi j}{N})| > 0$ for all $j = 0, \ldots, N-1$ or $|f(\frac{2\pi j}{N})| > 0$ for all $j = 1, \ldots, N$, then we use
\[
M_{N}(|f|^2, C_{N}^{(1)}) := (C_{N}^{(1)})^t \text{ diag} \left( |f(\frac{\pi j}{N})|^2 \right)_{j=0}^{N-1} C_{N}^{(1)},
\]
\[
M_{N}(|f|^2, S_{N}^{(1)}) := (S_{N}^{(1)})^t \text{ diag} \left( |f(\frac{\pi j}{N})|^2 \right)_{j=1}^{N} S_{N}^{(1)}
\]
as preconditioners, respectively.

4.3 Preconditioners for doubly symmetric block Toeplitz matrices with Toeplitz blocks

Finally, the generalization of our results to doubly symmetric block-Toeplitz systems with Toeplitz blocks is straightforward. We consider systems of linear equations
\[
A_{M,N} x = b,
\]
where \( A_{M,N} \) denotes a positive definite doubly symmetric block-Toeplitz matrix with Toeplitz blocks (BTTB matrix), i.e.
\[
A_{M,N} := (A_{r,s})_{r,s=0}^{M-1} \quad \text{with} \quad A_r := (a_{r,j-k})_{j,k=0}^{N-1}
\]
and \( a_{r,j} = a_{|r|,|j|} \). We assume that the matrices \( A_{M,N} \) are generated by a real-valued \( 2\pi \)-periodic continuous even function in two variables, i.e.
\[
a_{j,k} := \frac{1}{4\pi^2} \int_0^{2\pi} \int_0^{2\pi} \varphi(s, t) e^{-i(s^j + t^k)} \, ds \, dt.
\]
Lemma 4.1 can be extended to BTTB matrices as follows:
\[
A_{M,N} = (C^H_M \otimes C^H_N)' D_1 (C^H_M \otimes C^H_N) + (S^H_M \otimes C^H_N)' D_2 (S^H_M \otimes C^H_N)
\]
\[
+ (C^H_M \otimes S^H_N)' D_3 (C^H_M \otimes S^H_N) + (S^H_M \otimes S^H_N)' D_4 (S^H_M \otimes S^H_N)
\]
with
\[
D_1 := \text{diag} \left( \text{col}(\tilde{a}_{r,j})_{j,r=0}^{N-1,M-1} \right), \quad D_2 := \text{diag} \left( \text{col}(\tilde{a}_{r,j})_{j,r=0}^{N-1,M-1} \right),
\]
\[
D_3 := \text{diag} \left( \text{col}(\tilde{a}_{r,j})_{j,r=0}^{N,M-1} \right), \quad D_4 := \text{diag} \left( \text{col}(\tilde{a}_{r,j})_{j,r=0}^{N,M-1} \right),
\]
\[
(\tilde{a}_{r,j})_{j,r=0}^{N,M} := \tilde{C}_{M+1}^l \left( (a_{r,j})_{j,r=0}^{N,M} \right) (\tilde{C}_{N+1}^l)',
\]
\( a_{r,N} := 0 \ (r = 0, \ldots, M) \) and \( a_{M,j} := 0 \ (j = 0, \ldots, N) \). Here \( \text{col}: \mathbb{R}^{N,M} \to \mathbb{R}^{MN} \) is defined by
\[
\text{col} (x_{j,k})_{j,k=0}^{N-1,M-1} := (x_r)_{r=0}^{MN-1} \quad \text{with} \quad x_{N+j} := x_{j,k}.
\]
Consequently, the multiplication of a vector with a BTTB matrix requires only \( \mathcal{O}(MN \log(MN)) \) arithmetical operations. For details see [25]. We define our so-called “level-2” preconditioners by
\[
M_N(\varphi, C_M^H \otimes C_N^H) := (C_M^H \otimes C_N^H)' \text{diag}(\varphi(J_{M,N}^{\pi}))_{j,k=0}^{N-1,M-1} \times (C_M^H \otimes C_N^H),
\]
\[
M_N(\varphi, S_M^H \otimes S_N^H) := (S_M^H \otimes S_N^H)' \text{diag}(\varphi(J_{M,N}^{\pi}))_{j,k=0}^{N,M-1} \times (S_M^H \otimes S_N^H).
\]
Using the same arguments as in the Remark 4.2, we see that our PCG-method requires per iteration step only \( MN \) multiplications more than the conventional CG-method.
5 Numerical Examples

In this section, we show the efficiency of our new preconditioning technique by various numerical examples. The fast computation of the preconditioners and the PCG-method were implemented in MATLAB, where the C-programs for the fast trigonometric transforms were included by cmex. The algorithms were tested on a Sun SPARCstation 20.

As transform length we choose $N = 2^n$ and as right-hand side $b$ of (2.1) the vector consisting of $N$ entries “1”. The PCG-method started with the zero vector and stopped if $||r^{(j)}||_2/||r^{(0)}||_2 < 10^{-7}$, where $r^{(j)}$ denotes the residual vector after $j$ iterations.

It is remarkable, that the number of iterations also depends on roundoff-errors. Computation by the PCG-method which incorporates the FFT for the fast matrix-vector multiplications leads to another number of iteration steps than the same computation without FFT-techniques, i.e. with straightforward matrix-vector multiplications.

We begin with Hermitian ill-conditioned Toeplitz matrices $A_N(f)$ arising from the generating function

i) $f(x) = (x/2 - \pi/4)^4 \quad (x \in [0, 2\pi]).$

The second column of Table 5.1 shows the number of iterations of the CG-method without preconditioning. The columns 3 and 4 contain the numbers of iterations of the PCG-method with the optimal preconditioner $M_N^O(F_N)$ given by (1.3) and with our preconditioner $M_N(f, F_N)$ defined by (2.5) with $w := \pi/N$, respectively.

<table>
<thead>
<tr>
<th>$n$</th>
<th>$I_N$</th>
<th>$M_N^O(F_N)$</th>
<th>$M_N(f, F_N)$</th>
</tr>
</thead>
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<tr>
<td>4</td>
<td>26</td>
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<td>11</td>
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<tr>
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<td>2220</td>
<td>32</td>
</tr>
</tbody>
</table>

Next, we consider symmetric Toeplitz matrices $A_N$. We compare the Strang-type-preconditioners (4.4), our preconditioners (2.5) and (4.5) and the optimal trigonometric preconditioners defined by
DCT–II: \[ M_N^{(1)}(C_N^{(1)}) := (C_N^{(1)})^t \delta(C_N^{(1)} A_N(C_N^{(1)})^t) C_N^{(1)}, \]

DST–II: \[ M_N^{(1)}(S_N^{(1)}) := (S_N^{(1)})^t \delta(S_N^{(1)} A_N(S_N^{(1)})^t) S_N^{(1)}. \]

See for example [8, 12, 24]. Our test matrices correspond to the following generating functions:

ii) (see [26]): \( f(x) := (x^2 - 1)^2 \ (x \in [-\pi, \pi]) \).

In (2.5), we set \( w := \pi / N \).

iii) (see [26, 9, 10]): \( f(x) := x^4 \ (x \in [-\pi, \pi]) \).

In (2.5), we set \( w := \pi / N \).

The Tables 5.2 and 5.3 present the number of iteration steps for different preconditioners. The asterisk emphasizes that the corresponding preconditioners are not positive definite. Our new preconditioners lead to the best results. Compare also with [26, 9, 10]. Note that by the Remark 4.2, our PCG–method requires per iteration step only few arithmetical operations more than the conventional CG–method.

### Table 5.2: \( f(x) = (x^2 - 1)^2 \ (x \in [-\pi, \pi]) \)

<table>
<thead>
<tr>
<th>( n )</th>
<th>( I_N )</th>
<th>( C_N^{(1)} )</th>
<th>( S_N^{(1)} )</th>
<th>( C_N^{(1)} )</th>
<th>( S_N^{(1)} )</th>
<th>( F_N )</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>25</td>
<td>9*</td>
<td>8*</td>
<td>17</td>
<td>10</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>69</td>
<td>9*</td>
<td>8*</td>
<td>21</td>
<td>11</td>
<td>5</td>
</tr>
<tr>
<td>7</td>
<td>190</td>
<td>10*</td>
<td>10*</td>
<td>26</td>
<td>14</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>457</td>
<td>10*</td>
<td>10*</td>
<td>33</td>
<td>16</td>
<td>8</td>
</tr>
<tr>
<td>9</td>
<td>&gt;1000</td>
<td>11</td>
<td>9</td>
<td>43</td>
<td>19</td>
<td>9</td>
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<tr>
<td>10</td>
<td>&gt;1000</td>
<td>10*</td>
<td>10*</td>
<td>59</td>
<td>24</td>
<td>7</td>
</tr>
</tbody>
</table>

Our next test is related to non–Hermitian Toeplitz systems. As generating function of \( A_N(f) \) we choose

iv) (see [24]): \( f(x) = x^2 e^{i\pi} \ (x \in [-\pi, \pi]) \).

Then, the matrices \( A_N(f) \) have real entries such that we restrict our attention to trigonometric preconditioners. Table 5.5 compares the PCG–method applied to the normal equation (4.6) with

- the optimal preconditioner of \( A_N'(f) A_N(f) \) (see [24])
  \[ M_N^{(1)} := O_N \delta(O_N A_N'(f) A_N(f) O_N') O_N, \]

- the optimal preconditioner of \( A_N(|f|^2) \)
  \[ M_N^{(2)} := O_N \delta(O_N A_N(|f|^2) O_N') O_N, \]
Table 5.3: \( f(x) = x^4 \ (x \in [-\pi, \pi]) \)

<table>
<thead>
<tr>
<th>( n )</th>
<th>( I_N )</th>
<th>( C^H_N )</th>
<th>( S^H_N )</th>
<th>( C^H_N )</th>
<th>( S^H_N )</th>
<th>( S^H_N )</th>
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<tr>
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<td>15°</td>
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<tr>
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</tr>
</tbody>
</table>

- the Strang-type-preconditioner \( M_N(S_N f, O_N) \) and our preconditioner \( M_N(|f|^2, S^H_N) \) defined by (4.7).

Finally, let us turn to BTTB matrices \( A_{N,N} \). In our two examples, the matrices \( A_{N,N} \) are generated by the functions

v) (see [22]): \( \varphi(s, t) = s^2 t^4 \) and \( \psi(s, t) = (s^2 + t^2)^2 \ (s, t \in [-\pi, \pi]) \).

Both matrices are ill-conditioned and the CG-method without preconditioning, with Strang-type-preconditioning or with optimal trigonometric preconditioning converges very slow (see [22, 25]). Our preconditioning determined by (4.7) leads to the number of iterations in Table 5.5. Again, our PCG-method requires per iteration step only few arithmetical operations more than the conventional CG-method.

REFERENCES


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<th>191</th>
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<th>0.0001</th>
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<td>22</td>
<td>84</td>
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<td>5</td>
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</table>

![Table](image)

The table shows values for different variables, with columns and rows indicating specific data points.
Table 5.5: $\varphi(s,t) = s^2 t^4$ and $\psi(s,t) = (s^2 + t^2)^2$ ($s, t \in [-\pi, \pi]$)

<table>
<thead>
<tr>
<th>$N$</th>
<th>$M_N(\varphi, S_N^{II} \otimes S_N^{II})$</th>
<th>$M_N(\psi, S_N^{II} \otimes S_N^{II})$</th>
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</tbody>
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