On the decay of the off-diagonal singular values in cyclic reduction∗

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Abstract
It was recently observed in [9] that the singular values of the off-diagonal blocks of the matrix sequences generated by the Cyclic Reduction algorithm decay exponentially. This property was used to solve, with a higher efficiency, certain quadratic matrix equations encountered in the analysis of queuing models. In this paper, we provide a sharp theoretical bound to the basis of this exponential decay together with a tool for its estimation based on a rational interpolation problem. Applications to solving \( n \times n \) block tridiagonal block Toeplitz systems with \( n \times n \) semiseparable blocks and certain generalized Sylvester equations in \( O(n^2 \log n) \) arithmetic operations are shown.

Keywords: Cyclic reduction, quasiseparable matrices, rational interpolation, Sylvester equations, exponential decay, block tridiagonal systems.

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1 Introduction
Cyclic reduction, CR for short, is an algorithm originally introduced by G. H. Golub and R. W. Hockney in [17] for the solution of certain block tridiagonal linear systems coming from the finite difference discretization of elliptic PDEs. It has been later generalized and extended to other contexts, like for instance to the solution of polynomial matrix equations, and has been proven to be a successful method for solving a large class of queuing problems and infinite Markov Chains. We refer the reader to the books [8], [7] and to the survey paper [10] for more details and for the many references to the literature.

Given three \( m \times m \) matrices \( A_{-1}, A_0, A_1 \), and a positive integer \( n \) consider the block tridiagonal block Toeplitz matrix \( A_n = \text{trid}_n(A_{-1}, A_0, A_1) \) having block-size \( n \) where \( A_0 \) is on the main diagonal while \( A_{-1} \) is in the lower diagonal and \( A_1 \) in the upper diagonal. For a vector \( b \in \mathbb{R}^{mn} \), consider the system \( A_n x = b \). Roughly speaking, CR generates three sequences of \( m \times m \) matrices \( A_{-1}^{(h)}, A_0^{(h)} \) and \( A_1^{(h)} \), for \( h = 0, 1, \ldots \), with \( A_i^{(0)} = A_i, i = -1, 0, 1 \), and a sequence of systems \( A_{n_h} x^{(h)} = b^{(h)} \), \( A_{n_h} = \text{trid}_{n_h}(A_{-1}^{(h)}, A_0^{(h)}, A_1^{(h)}) \), where \( n_h = \lfloor n_h-1/2 \rfloor \) and \( x^{(h)} \) is a subvector of \( x \). This way, solving a block tridiagonal block-Toeplitz system of block size \( n \) is reduced to solving a block tridiagonal block Toeplitz system of size \( [n/2] \). The computation of \( A_i^{(h)} \) given

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exponential decay of the singular values is still evident with a basis of the exponential much
larger or even infinite systems, as well as matrix equations of the kind $A_{-1} + A_0 X + A_1 X^2 = 0$,
typically encountered in the analysis of queuing problems [21], where the unknown is the $m \times m$
matrix $X$ and a solution of spectral radius at most 1 is sought.

In short, the three sequences $A_i^{(h)}$, $i = -1, 0, 1$, which are related to the Schur complements
case, the matrix polynomial $z$ increasing order, then
values of the off-diagonal blocks of
numerical experiments show the high effectiveness of this approach.

Here we assume that all the matrices $\tilde{A}_i^{(h)}$ generated by the recursion are invertible so that CR
can be carried out with no breakdown. This assumption is generally satisfied in the applications.
In many cases of great interest, encountered for instance in the analysis of bi-dimensional
random walks, queuing models, network analysis [21], [24], [23], [18], [20], and finite differences
discretization of elliptic PDEs [13], the blocks $A$, $A_1$ are tridiagonal or, more generally,
banded matrices. This has raised great attention to the computational analysis of this case.
The additional tridiagonal structure makes it much cheaper to perform the first steps of CR where
the computational cost drops from $O(n^3)$ to $O(n)$ ops. However, after a few steps, the sparse
structure of the initial blocks is lost and one has to deal with full, apparently unstructured matrices $A_i^{(h)}$, $i = -1, 0, 1$.

Recently, in [0], it has been observed that if $A_{-1}$, $A_0$ and $A_1$ are tridiagonal, then the
matrices $\tilde{A}_i^{(h)}$, even if dense, numerically maintain a property of quasi-separability. That is, their
submatrices contained in the strict upper triangular part or in the strict lower triangular part,
called off-diagonal submatrices, have a “small” numerical rank. More formally, it has been proved
that if $\sigma_{i,h}$ are the singular values of any off-diagonal submatrix of, say, $A_0^{(h)}$, ordered in
non-increasing order, then $\sigma_{i,h} \leq \gamma t^{i/2}$ for some small $\gamma > 0$ and for some $0 < t < 1$. The value of $t$
is such that the matrix $z^2 A_1 + z A_0 + A_{-1}$ is invertible for any complex $z$ such that $t < |z| < t^{-1}$.

The analysis of [0] provides a theoretical explanation of an observed computational property
which enables one to implement CR with a high computational efficiency by relying on the
properties of quasiseparable matrices [27], [28]. In fact, an efficient implementation of CR has
been given based on the software library [11] of hierarchical quasiseparable matrices, and the
numerical experiments show the high effectiveness of this approach.

However, the results of [0] provide an under estimate of the decay properties of the singular
values of the off-diagonal blocks of $A_{-1}^{(h)}$, $A_0^{(h)}$ and $A_1^{(h)}$. In fact, it turns out that, even in the
cases where the matrix polynomial $z^2 A_1 + z A_0 + A_{-1}$ is singular at some point just outside a
thin annulus $A_t = \{ z \in \mathbb{C} : \ t \leq |z| \leq t^{-1} \}$ obtained with some $t$ very close to 1, the observed
exponential decay of the singular values is still evident with a basis of the exponential much
smaller than the given theoretical bound $t$.

A typical example is given by the discrete Laplacian matrix where $A_0 = \text{trid}(-1, 4, -1)$,
$A_{-1} = A_1 = -I$ so that $t = 1 - 1/(n + 1) + O(1/(n + 1)^2)$. In this case, for moderately large
values of \( n \), the bound \( t^j \) is still close to 1 for values of \( j \) as large as \( n \). As a result, the plot of the upper bounds to the singular values would be an almost horizontal line. On the other hand from the numerical experiments it turns out that the decay of the singular values is still exponential despite the width of the annulus collapses to zero, and the basis of the exponential is much less than \( t \) and almost independent of \( n \).

With the tools introduced in this paper, we can capture this property as shown in Figure 1 where the decay of the off-diagonal singular values of the matrix \( H_0 = \lim_{h \to 0} A_0^{(h)} \), together with their theoretical upper bounds, are displayed.

In fact, in this paper we provide a different theoretical explanation of the exponential decay of the singular values which relies on an unpublished result of B. Beckermann [3] where the decay of certain singular values associated with some Krylov matrix is expressed in terms of the accuracy of a rational function approximation problem.

This analysis leads to a fast algorithm, which we call *quasiseparable CR* (QCR for short), for solving an \( n \times n \) block tridiagonal block Toeplitz system, where the matrix \( \text{trid}_m(B, A, C) \) has quasiseparable, say tridiagonal, \( m \times m \) blocks \( A, B, C \). The cost of the algorithm is \( O(m n \log m + m \log^2 m \log n) \) ops which reduces to \( O(n^2 \log n) \) ops for \( m = n \). This cost is comparable with that of the fast Poisson solvers [14, Sect. 4.8.4], which apply to the case where \( A = \text{trid}_n(-1, 4, 1), B = C = -I \). Unlike the latter algorithms, quasiseparable CR covers a wider and more general class of cases.

We show also an application of QCR to solving a generalized Sylvester equation of the kind

\[
\sum_{i=1}^{k} A_i X B_i = C
\]

for given matrices \( A_i, B_i \) and \( C \) of compatible sizes, in the case where \( A_i \) are tridiagonal Toeplitz,
and $B_i$ are quasiseparable matrices. In fact, in this case, the problem is reduced to solve a block tridiagonal block Toeplitz system with quasiseparable blocks. The cost of the solution is again $O(n^2 \log n)$ ops where, for simplicity, we assume that all the matrices involved are $n \times n$.

Decay properties of the off-diagonal blocks of matrix functions have been recently received much attention. In particular, in the paper by M. Benzi and P. Boito and N. Razouk the decay properties of spectral projectors associated with large and sparse Hermitian matrices are investigated. In D. Kressner and A. Susnjara prove a priori bounds for the numerical rank of the off-diagonal blocks of spectral projectors—associated with symmetric banded matrices—by using the best rational approximant of the sign function. In M. Benzi and P. Boito extend previous results on the exponential off-diagonal decay of the entries of analytic functions of banded and sparse matrices to the case where the matrix entries are elements of a $C^*$-algebra. M. Benzi and V. Simoncini find decay bounds for completely monotonic functions of matrices which are the Kronecker sum of banded or sparse matrices. While C. Canuto, V. Simoncini and M. Verani analyze the decay pattern of the inverses of banded matrices of the form $I \otimes M + M \otimes I$ where $M$ is tridiagonal, symmetric and positive definite. In S. Chandrasekaran, P. Dewilde, M. Gu, and N. Somasunderam analyze the numerical rank of the off-diagonal blocks in the Schur complements of block tridiagonal block Toeplitz systems discretizing bi-dimensional elliptic equations.

The paper is organized as follows. In Section 2 we provide some preliminary results including the main properties of CR, its functional interpretation, and the definitions of $k$-quasiseparable matrices and of hierarchical $k$-quasiseparable matrices. Section 3 concerns the analysis of the properties of the matrix coefficients in the Laurent expansion of the matrix function $\psi(z) = \varphi(z)^{-1}$, where $\varphi(z) = z^{-1}A_{-1} + A_0 + zA_1$. In fact, this matrix function captures the structural and computational properties of CR. Its domain of analyticity is the annulus $\mathcal{A}_t$ whose width has been used in to prove the exponential decay. The main result of this section is Lemma 3.2 where we show that any off-diagonal block of $\psi(z)$ can be written as the sum of 4 terms; each term is the product of a Krylov matrix and of a transposed Krylov matrix.

In Section 4—we rely on a result by B. Beckermann—we provide a bound to the singular values of a matrix which satisfies a suitable displacement equation. Then we apply this result to find sharp bounds to the singular values of the off-diagonal blocks of $\psi^{(h)}(z)$ and we extend these bounds to the block $A^{(h)}_i$ and to the limit value $\lim_{h} \psi^{(h)}(z) = A^{(\infty)}_0$. The main results of this section are given in Theorems 4.6 and 4.7.

Section 5 deals with the experimental validation of the theoretical bounds to the decay. In Section 6 we show applications of the quasiseparable CR to solving block tridiagonal block Toeplitz systems and to solving certain generalized Sylvester equations. We report also the results of some numerical experiments where the above applications are tested. Finally, Section 7 draws the conclusions.

2 Some preliminaries

Throughout, $\mathbb{Z}$ and $\mathbb{N}$ denote the set of relative integers and of natural numbers, respectively, while $\mathbb{R}$ and $\mathbb{C}$ denote the complex and the real field, respectively. We recall the fundamental properties of CR and of quasiseparable matrices.

Cyclic reduction can be formulated in functional form in terms of two matrix Laurent series $\varphi(z)$ and $\psi(z)$, namely,

$$\varphi(z) = z^{-1}A_{-1} + A_0 + zA_1, \quad \psi(z) = \varphi(z)^{-1},$$

where $\psi(z)$ is defined in the set where $\varphi(z)$ is invertible. Here and hereafter, we assume that $\varphi(z)$
is invertible in the annulus $A_t = \{ z \in \mathbb{C} : t \leq |z| \leq t^{-1} \}$ for some $0 < t < 1$. This assumption is generally verified in the applications. In certain cases, by means of scaling the matrices $A_1$ and $A_{-1}$ by suitable constants $\alpha$ and $\alpha^{-1}$, respectively, one can meet this assumption. Throughout we denote $T = A_1$ the unit circle in the complex plane.

We recall the following property which is fundamental for our analysis, see for instance [8] and [9].

**Proposition 2.1.** Let $A_{-1}, A_0, A_1$ be $n \times n$ matrices such that CR can be carried out. Define $\varphi_h(z) = z^{-1}A_{-1}^{(h)} + A_0^{(h)} + zA_1^{(h)}$, where $A_i^{(h)}$ are the matrices generated by $[1]$, and set $\psi_h(z) = \varphi_h(z)^{-1}$. Then

$$
\psi(h)(z^{2^k}) = \frac{1}{2^k} \sum_{j=1}^{2^k} \psi(\xi_{2^k}, z)
$$

where $\xi_{2^k}$ is a primitive $2^k$-th root of the unity.

The following definitions are fundamental to formalize the fast decay of the singular values of the off-diagonal submatrices generated by CR. We say that an $m \times m$ matrix $A$ is $k$-quasiseparable if all the submatrices contained in the strict upper triangular part or in the strict lower triangular part have rank at most $k$ and there exists at least one submatrix with rank $k$. We say also that $k$ is the quasiseparable rank of $A$.

We say that $A$ is hierarchically $k$-quasiseparable if either $m \leq k$ or there exists a $2 \times 2$ block partitioning of the matrix such that the diagonal blocks are square and have size $\lfloor \frac{m}{2} \rfloor$ and $\lceil \frac{m}{2} \rceil$, respectively, the off-diagonal blocks have rank at most $k$ and the diagonal blocks are hierarchically $k$-quasiseparable. Moreover, in this recursive partitioning there exists an off-diagonal submatrix of rank exactly $k$.

This partitioning leads to the simplest hierarchical representation, known in the literature as hierarchically off-diagonal low rank (HODLR), which is the one exploited in [9] for speeding up CR.

The following result states that if the singular values of the off-diagonal blocks of $A$ decay fast, then $A$ is close to a hierarchical quasiseparable matrix. That is, for a relatively small $k$ there is a perturbation $\delta A$ of small norm such that $A + \delta A$ is hierarchically $k$-quasiseparable.

**Theorem 2.2.** Let $f(l)$ be a function over the positive integers, and let $A \in \mathbb{C}^{m \times m}$ be a matrix such that $\sigma_j(B) \leq f(l)$ for every off-diagonal block $B$ in $A$. Then, for any $l$ there exists a perturbation matrix $\delta A$ such that $A + \delta A$ is hierarchical quasiseparable of rank at most $l$ and $\|\delta A\|_2 \leq f(l) \cdot \log_2 m$.

**Proof.** First, recall that if the nonzero singular values of a matrix $B$ are $\sigma_1 \geq \sigma_2 \geq \ldots \geq \sigma_k$ then, for any $j \leq k$ we may write $B$ as a matrix of rank $j$ plus a perturbation $\delta B$ such that $\|\delta B\|_2 = \sigma_{j+1}$. Now consider an HODLR like partitioning of $A$ with minimal blocks of dimension $l$. Notice that the depth of this recursive partition is $\sigma = \lceil \log_2 \left( \frac{m}{T} \right) \rceil$. This way, for each off-diagonal block $B$ of this partitioning and for any integer $j$, there exists a perturbation matrix that makes this block of rank $j$. The $2$-norm of this perturbation is equal to $\sigma_{j+1}(B) \leq f(j+1)$. We may form the matrix $\delta A$ which collects all these perturbations of each off-diagonal block of the above partitioning. This way, if $j = l - 1$, the off-diagonal blocks of $A + \delta A$ have rank at most $l$. We can now show that $\|\delta A\|_2 \leq f(l) \cdot \log_2 m$. We have

$$
\delta A = \sum_{i=0}^\sigma \delta A_i, \quad \sigma \leq \log_2 m,
$$
where $\delta A_i$ is the correction obtained by putting together all the blocks at level $i$ of subdivision, that is,

$$
\delta A_0 = \begin{bmatrix}
0 & \delta X^{(0)}_1 \\
\delta X^{(0)}_2 & 0
\end{bmatrix}, \quad \delta A_1 = \begin{bmatrix}
0 & \delta X^{(1)}_1 \\
\delta X^{(1)}_2 & 0
\end{bmatrix}, \quad \ldots
$$

Since the summands are just permutations of block diagonal matrices their 2-norm is the maximum of the 2-norms of the (block) diagonal entries, and this gives the desired bound.

Thus, our aim is to prove that the matrix function $\psi^{(h)}(z)$ defined in Proposition 2.1 has off-diagonal blocks with singular values which decay exponentially to zero so that the assumptions of Theorem 2.2 are satisfied with $f(l) = e^{-\alpha l}$ for some positive $\alpha$. This decay property is then extended to $\varphi_{h}(z)$ by inversion and finally to the blocks $A^{(h)}_{-1}, A^{(h)}_{0}, A^{(h)}_{1}$ by means of interpolation. More details on this technique are given in [9].

The estimates of the parameter $\alpha$ given in the paper [9] depend on the value $t$ which defines the domain $\mathcal{A}_t$ of invertibility of the matrix $\varphi(z)$. If $t$ gets close to 1, then $\alpha$ takes values close to 0, and the theoretical bound of the exponential decay loses its sharpness. Here, we introduce a different analysis which better fits with the decay observed in the numerical experiments.

We define the following class of problems for which the matrices $A^{(h)}_i$, $i = -1, 0, 1$ generated by CR through (1) have the exponential decay of the singular values in their off-diagonal blocks at any step $h$ of the iteration.

**Definition 2.3.** Let $\varphi(z) = z^{-1}A_{-1} + A_0 + zA_1$, where $A_{-1}, A_0, A_1$ are $m \times m$ matrices with entries in $\mathbb{C}$, be such that CR can be applied with no breakdown by means of (1). Let $f(l)$ be a positive function in $l^1(\mathbb{N})$. We say that $\varphi(z)$ is $f$-decaying-quasiseparable if, for all $h \in \mathbb{N}$, $\forall z \in \mathbb{T}$ and for every off-diagonal block $\tilde{C}^{(h)}(z)$ of $\psi^{(h)}(z)$, we have

$$
\sigma_l(\tilde{C}^{(h)}(z)) \leq \|\psi^{(h)}(z)\|_2 \cdot f(l),
$$

where $\sigma_l(\tilde{C}^{(h)}(z))$ denotes the $l$-th singular value of the matrix $\tilde{C}^{(h)}(z)$. We define the set of such matrix functions $\varphi(z)$ as $\text{DQ}(f)$.

## 3 Laurent coefficients of an off-diagonal block

In this section, we consider the matrix Laurent series expansion of $\psi(z)$, that is, $\psi(z) = \sum_{i=-\infty}^{+\infty} z^i H_i$ for $z \in \mathcal{A}_t$, which exists and is convergent since $\psi(z)$ is analytic in the domain $\mathcal{A}_t$ where $\varphi(z)$ is analytic and non-singular. Then we will analyze the properties of the coefficients of an off-diagonal block of this Laurent series.

We define the eigenvalues of $\varphi(z)$ as the roots of the polynomial $p(z) = \det(A_{-1} + zA_0 + z^2A_1)$. Observe that if $\det A_1 \neq 0$ the polynomial $p(z)$ has degree $d = 2m$ so that there are $2m$ roots. If, on the other hand, $\det A_1 = 0$ then $d < 2m$ and for this reason, we add to the $d$ roots of $p(z)$ other $2m - d$ roots at the infinity. In this way we can say that $\varphi(z)$ has always $2m$ eigenvalues including possible eigenvalues at the infinity.

Here we assume that the eigenvalues $\xi_i$, $i = 1, \ldots, 2m$ of $\varphi(z)$ satisfy the balanced splitting property with respect to the unit circle

$$
|\xi_1| \leq \ldots \leq |\xi_m| < t < 1 < t^{-1} < |\xi_{m+1}| \leq \ldots \leq |\xi_{2m}|. \tag{2}
$$
We call $t$ the radius of the splitting. The splitting property \( [2] \) is needed to guarantee the applicability of CR and that the convergence to zero of the blocks $A_{h-1}^h$ and $A_1^h$ is doubly exponential \([8]\).

Consider the following partitioning of $\psi(z)$ and $\varphi(z)$

$$
\varphi(z) = \begin{pmatrix} A(z) & B(z) \\ C(z) & D(z) \end{pmatrix}, \quad \psi(z) = \begin{pmatrix} \tilde{A}(z) & \tilde{B}(z) \\ \tilde{C}(z) & \tilde{D}(z) \end{pmatrix} = \begin{pmatrix} S_D(z)^{-1} & * \\ -D(z)^{-1}C(z)SD(z)^{-1} & * \end{pmatrix},
$$

where the diagonal blocks are square, $S_D(z) = A(z) - B(z)D(z)^{-1}C(z)$ is the Schur complement of $D(z)$, and $*$ denotes blocks which are not relevant for our analysis.

Moreover, suppose that the splitting \([2]\) holds also for the eigenvalues of $D(z)$—this is true for problems from stochastic models which are ruled by M-matrices—and assume that the matrix coefficients $A_i$ have quasireducible rank $k$ for $i = -1,0,1$. These hypotheses are always satisfied for a large class of important problems like not-null recurrent Quasi Birth-Death problems (QBDs) with banded blocks, up to rescaling the coefficients \([8]\). This guarantees that the matrix functions $\varphi(z)$ and $D(z)$ are invertible in the annulus $A_{k}$ for some $t < 1$.

Observe that, since the off-diagonal blocks of $A_i$ have rank at most $k$ for $i = -1,0,1$, then any off-diagonal block $C(z)$ of $\varphi(z)$ can be written as

$$
C(z) = z^{-1}U_{-1}V_{-1}^t + U_0V_0^t + zU_1V_1^t, \quad \|U_i\| = \|A_i\|, \quad \|V_i\| = 1,
$$

where $U_i$ and $V_i$ have $k$ columns and the superscript $t$ denotes transposition.

Defining

$$
U = \begin{bmatrix} U_{-1} & U_0 & U_1 \end{bmatrix}, \quad V(z) = \begin{bmatrix} z^{-1}V_{-1} & V_0 & zV_1 \end{bmatrix},
$$

we can write

$$
\tilde{C}(z) = -\tilde{U}(z)\tilde{V}(z)^t, \quad \tilde{U}(z) = D(z)^{-1}U \quad \text{and} \quad \tilde{V}(z) = S_D(z)^{-1}V(z).
$$

Observe that $S_D(z)^{-1}$ is the upper left diagonal block of $\psi(z)$. This gives us a crucial information on the coefficients of the matrix Laurent series expansion of $D(z)^{-1}$ and $S_D(z)^{-1}$. In order to perform this analysis we have to recall a general result which provides an explicit expression of the coefficients $H_i$ of the Laurent expansion of $\psi(z)$.

**Theorem 3.1** (Part of Theorem 3.20 in \([8]\)). Let $\varphi(z) = z^{-1}A_{-1} + A_0 + zA_1$ with $A_i \in \mathbb{R}^{m \times m}$, $i = -1,0,1$ and assume that the eigenvalues $\xi_i$, $i = 1,\ldots,2m$ of $\varphi(z)$ satisfy \([2]\). Moreover suppose that there exist $R$ and $\tilde{R}$ with spectral radius less than $1$ which solve the matrix equations

$$
A_1 + XA_0 + X^2A_{-1} = 0, \quad \text{(3)}
$$

$$
X^2A_1 + XA_0 + A_{-1} = 0, \quad \text{(4)}
$$

respectively. Then there exist $G$ and $\tilde{G}$ solutions of the reversed matrix equations

$$
A_1X^2 + A_0X + A_{-1} = 0, \quad \text{(5)}
$$

$$
A_1 + A_0X + A_{-1}X^2 = 0, \quad \text{(6)}
$$

respectively, with spectral radius less than $1$. Moreover, expanding $\varphi(z)^{-1} = \sum_{j=-\infty}^{+\infty} z^j H_j$ yields

$$
H_j = \begin{cases} 
H_0\tilde{R}^{-j} = G^{-j}H_0 & j \leq 0, \\
H_0R^j = \tilde{G}^jH_0 & j \geq 0,
\end{cases} \quad G = H_{-1}H_0^{-1}, \quad \tilde{G} = H_1H_0^{-1}, \quad R = H_0^{-1}H_1, \quad \tilde{R} = H_0^{-1}H_{-1}.
$$

The spectrum of $G$ and $\tilde{R}$ is formed by the eigenvalues of $\varphi(z)$ inside the unit disc, the spectrum of $\tilde{G}$ and $R$ is formed by the reciprocals of the eigenvalues of $\varphi(z)$ outside the unit disc.
This result, applied with \( \varphi(z) = D(z) \) and combined with what said previously, tells us that the Laurent coefficients of \( \tilde{U}(z) \) are of the form

\[
D^{-1}(z) = \sum_{j \in \mathbb{Z}} z^j H_{D,j}, \quad H_{D,j} = \begin{cases} G^{-j}_D H_{D,0} & j \leq 0, \\ \tilde{G}_D^j H_{D,0} & j \geq 0, \end{cases}
\]

where \( G_D \) and \( \tilde{G}_D \) are the solutions of the matrix equations associated with \( D(z) \) of the kind \(^5\) and

\[
S_D(z)^{-1} = \sum_{j \in \mathbb{Z}} z^j H_{S,j}, \quad H_{S,j} = \begin{cases} [I \ 0] H_0 \tilde{R}^{-j} [I \ 0]^t & j \leq 0, \\ [I \ 0] H_0 R^j [I \ 0]^t & j \geq 0, \end{cases}
\]

where the latter equation is obtained by applying Theorem 3.1 to the original matrix Laurent polynomial \( \varphi(z) \).

Consider the simpler case where \( k = 1 \) and the decomposition of each off-diagonal block \( C(z) \) of \( \varphi(z) \) can be written as \( C(z) = uu^t \) (a constant dyad). This is not restrictive since, in the other cases, we can write \( C(z) \) as a linear combination of at most 3\( k \) terms of the above form with coefficients \( z^j, j = -1, 0, 1 \).

In view of Theorem 3.1, for \( z \in \mathbb{T} \) we can write each off-diagonal block \( \tilde{C}(z) \) of \( \psi(z) \) as

\[
\tilde{C}(z) = \tilde{u}(z) \tilde{v}(z)^t, \quad \tilde{u}(z) = \sum_{j \geq 0} \tilde{G}_D^j H_{D,0} u z^j + \sum_{j < 0} G^{-j}_D H_{D,0} u z^j,
\]

where \( \tilde{v}(z) = S_D(z)^{-1} v \), the matrix function \( S_D(z)^{-1} \) is the inverse of the Schur complement of \( D(z) \) and \( \|v\|_2 \leq 1 \). Observe that the Laurent coefficients of \( \tilde{u}(z) \) corresponding to positive powers of \( z \) lie in the Krylov subspace \( K_j(G_D, H_{D,0} u) \), while the coefficients corresponding to the negative powers are in \( K_j(G_D, H_{D,0} u) \). Here we denote by \( K_j(A, v) \) the \( (j+1) \)-dimensional Krylov subspace

\[
K_j(A, v) = \text{span}(v, Av, A^2 v, \ldots, A^j v).
\]

Analogously we know that

\[
v^t S_D(z)^{-1} = \left( \sum_{j > 0} \tilde{v}^t H_{\psi,0} R^j z^j + \sum_{j < 0} \tilde{v}^t H_{\psi,0} \tilde{R}^{-j} z^j \right) \begin{bmatrix} I \\ 0 \end{bmatrix}, \quad \tilde{v} := \begin{bmatrix} v \\ 0 \end{bmatrix},
\]

therefore

\[
-\tilde{C}(z) = \left( \sum_{j > 0} \tilde{G}_D^j H_{D,0} u z^j + \sum_{j < 0} G^{-j}_D H_{D,0} u z^j \right) \cdot \left( \sum_{j > 0} \tilde{v}^t H_{\psi,0} R^j z^j + \sum_{j < 0} \tilde{v}^t H_{\psi,0} \tilde{R}^{-j} z^j \right) \begin{bmatrix} I \\ 0 \end{bmatrix}.
\]

Denoting by \( \tilde{C}^{(h)}(z^{2h}) \) the corresponding off-diagonal sub-block in \( \psi^{(h)} \), from Proposition 2.1 we have

\[
\tilde{C}^{(h)}(z^{2h}) = \frac{1}{2h} \sum_{j=1}^{2h} \tilde{C}(z^{j2h}).
\]
In the following, the matrices with columns of the form $A^jb$, for some matrix $A$ and a vector $b$, which we call Krylov matrices, will play an important role. We indicate a Krylov matrix with the notation

$$KM_n(A, b) := \begin{bmatrix} b \mid Ab \mid \ldots \mid A^{n-1}b \end{bmatrix}.$$ 

Moreover, we denote by $J$ the counter-identity matrix of appropriate size such that $[1, 2, \ldots, n]J = [n, n-1, \ldots, 1]$.

Relying on (7) we can prove the following result.

**Lemma 3.2.** If $C(z) = uv^t$, then $-\tilde{C}^{(h)}(z^2h)$ is the sum of the following four outer products:

$$-\tilde{C}^{(h)}(z^2h) = \begin{bmatrix} KM_{2h}(G_D, \hat{a}) \cdot KM_{2h}(\tilde{R}^t, \hat{b})^t \\
+z^{2h-1} \cdot KM_{2h}(\tilde{G}_D, \hat{a}) \cdot J \cdot KM_{2h}(R^t, b)^t \\
+z^{1-2h} \cdot KM_{2h}(G_D, a) \cdot J \cdot KM_{2h}(\tilde{R}^t, \tilde{b})^t \\
+ KM_{2h}(G_D, a) \cdot KM_{2h}(R^t, b)^t \end{bmatrix} \begin{bmatrix} t \\
0 \end{bmatrix},$$

where

$$a = \left( \sum_{s \in 2hZ \cap \mathbb{N}} z^{-s-1}G_D^{s+1} \right) H_{D,0}u, \quad b = \left( \sum_{s \in 2hZ \cap \mathbb{N}} z^{s+1}R^{s+1} \right) H_{\psi,0}\hat{v},$$

$$\hat{a} = \left( \sum_{s \in 2hZ \cap \mathbb{N}} z^s G_D^s \right) H_{D,0}u, \quad \hat{b} = \left( \sum_{s \in 2hZ \cap \mathbb{N}} z^{-s}R^s \right) H_{\psi,0}\hat{v}.$$

**Proof.** Thanks to (7) we may write $-\tilde{C}(z)$ as the sum of four outer products. By the linearity of (8) we can consider them separately. Take for example

$$\left( \sum_{j \geq 0} \tilde{G}^j_D H_{D,0}u z^j \right) \cdot \left( \sum_{j \geq 0} \tilde{v}^j H_{\psi,0} \tilde{R}^{-j} z^j \right) = \sum_{j \geq 0} \tilde{G}^j_D H_{D,0}u \tilde{v}^j H_{\psi,0} \sum_{s \geq 0} \tilde{R}^s z^{j-s},$$

where we have ignored $[1 \ 0]^t$ because it can be factored on the right. The block $\tilde{C}^{(h)}(z)$ of $\psi^{(h)}(z)$ corresponding to $\tilde{C}(z)$ in $\psi(z)$ verifies the relation $\tilde{C}^{(h)}(z^2h) = \frac{1}{2^h} \sum_{l=1}^{2^h} \tilde{C}(z^{2h}l)$ so that

$$\frac{1}{2^h} \sum_{l=1}^{2^h} \sum_{j \geq 0} \tilde{G}^j_D H_{D,0}u \tilde{v}^j H_{\psi,0} \sum_{s \geq 0} \tilde{R}^s z^{j-s} = \sum_{j \geq 0} \tilde{G}^j_D H_{D,0}u \tilde{v}^j H_{\psi,0} \sum_{s \in (2^hZ+j) \cap \mathbb{N}} \tilde{R}^s z^{j-s} = \sum_{j=0}^{2^h-1} \sum_{s \in (2^hZ+j) \cap \mathbb{N}} z^s G_D^s H_{D,0}u \tilde{v}^j H_{\psi,0} \sum_{s \in (2^hZ+j) \cap \mathbb{N}} \tilde{R}^s z^{j-s},$$

where $2^hZ + j := \{ s \in Z \mid s \equiv j \mod 2^h \}$. Observe that the $(j+1)$-st term of the previous sum is equal to the $j$-th term multiplied on the left by $z\tilde{G}_D$ and on the right by $z^{-1}\tilde{R}$. In particular we can rewrite it as

$$\left[ a \mid z\tilde{G}_D \cdot a \ldots \mid (z\tilde{G}_D)^{2h-1} \cdot a \right] \cdot \left[ \hat{b} \mid (z^{-1}\tilde{R})^t \cdot \hat{b} \mid \ldots \mid (z^{-1}\tilde{R})^{2h-1} \cdot \hat{b} \right]^t,$$
that is, $\mathcal{K}\mathcal{M}_{2b}(z\hat{G}_D, a) \cdot \mathcal{K}\mathcal{M}_{2b}(z^{-1}\hat{R}, \hat{b})$.

The variables $z$ in the above factors cancel out, and we obtain one of the addends in the statement of the theorem.

Then consider $\left(\sum_{j>0} \hat{G}_D^j H_{D,0} u z^j\right) \cdot \left(\sum_{j>0} \hat{t}^j H_{\psi,0} R^{-j} z^j\right)$ for which we arrive at the expression

$$\sum_{j=0}^{2h-1} \left(\sum_{s \in (2b^2 + j) \cap \mathbb{N}} z^s \hat{G}_D^j H_{D,0} u \hat{t}^j H_{\psi,0} \left(\sum_{s \in (2b^2 - j) \cap \mathbb{N}} R^s z^s\right)\right).$$

This time we have a product of the form

$$\left[ a \; z\hat{G}_D \cdot a \; \ldots \; (z\hat{G}_D)^{2h-1} \cdot a \right] \cdot \left[ (z\hat{R})^{2h-1} \cdot b \; \ldots \; (z\hat{R}) \cdot b \; b \right]^t,$$

that is $z^{2h-1} \cdot \mathcal{K}\mathcal{M}_{2b}(\hat{G}_D, a) \cdot J \cdot \mathcal{K}\mathcal{M}_{2b}(\hat{R}, \hat{b})$. The other two relations are obtained in a similar manner.

In the case $C(z) = z^s w^t$ with $s = -1, 1$ one can recover the same behavior just taking into account a shift in the powers of $z$ in (7) that modifies the powers of $z$ in the outer products accordingly.

4 Singular values and displacement rank

Lemma 3.2 provides a tool for analyzing the singular values decay of the off-diagonal block $\tilde{C}(b)(z)$ of $\tilde{C}(b)(z)$. In fact, these blocks can be written as the sum of few terms each of them is the product of two Krylov matrices, one of which is transposed.

The next step is to investigate the singular values of a product of this kind. In this analysis, we rely on the concept of displacement rank and on some result by B. Beckermann [3], of which we report the proof.

Definition 4.1. Given matrices $A, B, X \in \mathbb{C}^{m \times m}$ the displacement rank of $X$ with respect to the pair $(A, B)$ is defined as

$$\rho_{A,B}(X) = \text{rank}(AX - XB).$$

We need also to introduce the set $\mathcal{R}_{n,d}$ of rational functions over $\mathbb{C}$ where $n$ and $d$ are the degree of the numerator and of the denominator, respectively.

For a matrix $X$ with a small displacement rank it is possible to provide bounds on its singular values in terms of the optimal values of some Zolotarev problems [29] according to the following result of B. Beckermann [3].

Theorem 4.2. Let $X \in \mathbb{C}^{m \times m}$ and suppose that there exist two normal matrices $A, B \in \mathbb{C}^{m \times m}$ such that $\rho_{A,B}(X) = d$. Then, indicating with $E$ and $F$ the spectrum of $A$ and $B$ respectively, for the singular values $\sigma_i(X)$ of $X$ it holds:

$$\frac{\sigma_{1+l,i}(X)}{\|X\|_2} \leq Z_l(E, F) := \inf_{r(x) \in \mathcal{R}_{l,d}} \max_{r(x) \in \mathcal{R}_{l,d}} \|r(x)\| \min_{r(x) \in F} |r(x)|, \quad l = 1, 2, \ldots.$$

Proof. Consider $p(x) := \sum_{i=0}^l p_i x^i$ and $q(x) := \sum_{i=0}^l q_i x^i$ polynomials of degree $l$ and define $r(x) := \frac{p(x)}{q(x)}$. We prove that the matrix

$$\Delta := q(A)X p(B) - p(A)X q(B)$$
has rank at most $l \cdot d$. Without loss of generality we consider the case $d = 1$ and suppose $AX - XB = uv^t$. We can prove by induction that $A^kX - XB^k = \sum_{h=0}^{k-1} A^huv^t B^{k-1-h}$. For $k = 1$ the property trivially holds. For $k > 1$ one has:

$$A^kX = A^{k-1}XB + A^{k-1}uv^t = XB^k + \left( \sum_{h=0}^{k-2} A^huv^t B^{k-1-h} \right) B + A^{k-1}uv^t$$

$$= XB^k + \sum_{h=0}^{k-1} A^huv^t B^{k-1-h}.$$

Now, observe that

$$\Delta = q(A)Xp(B) - p(A)Xq(B) = \sum_{i \neq j} (q_ip_j - q_jp_i) (A^iXB^j - A^jXB^i),$$

and if $i > j$ (the other case is analogous)

$$A^iXB^j - A^jXB^i = A^j(A^{i-j}X - XB^{i-j})B^i = A^j \left( \sum_{h=0}^{i-j-1} A^huv^t B^{k-1-h} \right) B^j.$$

In particular all the addends involved in the expansion of $\Delta$ can be expressed as sum of dyads whose left vectors belong to the Krylov space $K_l(A, u)$ and so it has rank at most $l$.

Assume that $q(A)$ and $p(B)$ are invertible, define $Y := q(A)^{-1} \Delta p(B)^{-1}$ observe that $X - Y = r(A)Xr(B)^{-1}$ so that

$$\|X - Y\|_2 = \|r(A)Xr(B)^{-1}\|_2 \leq \|X\|_2 \max_{E,F} \left| r(x) \right| \max_{E,F} \left| r(x) \right|^{-1} = \|X\|_2 \max_{E,F} \left| r(x) \right| \min_{E,F} \left| r(x) \right|^{-1}.$$

Since $\sigma_{k+1}(X)$ coincides with the minimum of $\|X - W\|_2$ taken over all the matrices $W$ of rank $k$, and since rank($Y$) = rank($\Delta$) $\leq l \cdot d$, we find that

$$\sigma_{l+d+1}(X) \leq \|X - Y\|_2 \leq \|X\|_2 \max_{E,F} \left| r(x) \right| \min_{E,F} \left| r(x) \right|^{-1}.$$

Taking the infimum over the set of rational functions of degree $(d, d)$ completes the proof.

Note that the normality hypothesis can be relaxed by replacing it with the diagonalizability of $A$ and $B$. The price to pay is a larger constant depending on the conditioning of the eigenvector matrices as stated by the following

**Corollary 4.3.** Let $X \in \mathbb{C}^{m \times m}$ and suppose that there exist two diagonalizable matrices $A, B \in \mathbb{C}^{m \times m}$ such that $\Delta_{A,B}(X) = d$, that is $A = V_AD_AV_A^{-1}$, $B = V_BD_BV_B^{-1}$ with $D_A$ and $D_B$ diagonal matrices. Then, indicating with $E$ and $F$ the spectrum of $A$ and $B$ respectively, it holds:

$$\sigma_{l+d+1}(X) \leq Z_l(E,F) \cdot \|X\|_2 \cdot \kappa(V_A) \cdot \kappa(V_B)$$

where $\kappa(W) = \|W\|_2 \|W^{-1}\|_2$ denotes the spectral condition number of $W$.

The case where $E$ and $F$ are disjoint subsets of the real line, has been extensively studied by Zolotarev [29] who managed to provide explicit bounds for $Z_l(E,F)$. The result we are going to quote is adapted to our case and can be found in [15]. See also [1, 2, 22] for more classical references.
Theorem 4.4 (Zolotarev). Let \( \delta \in (0, 1) \), \( E := [-\infty, -\delta^{-1}] \cup [\delta^{-1}, +\infty] \) and \( F = [-\delta, \delta] \). Then

\[
Z_{2l}(E, F) \leq \frac{2\rho}{1 - 2\rho},
\]

where

\[
\rho := \exp \left( -\frac{\pi}{2} K \left( \sqrt{1 - \delta^2} \right) \right), \quad K(x) := \int_0^1 \frac{1}{\sqrt{(1 - t^2)(1 - x^2 t^2)}} dt.
\]

Moreover, if \( \delta \approx 1 \) then \( K(\delta^2) \approx \log \left( \frac{4}{\sqrt{1 - \delta^4}} \right) \) and \( K(\sqrt{1 - \delta^2}) \approx \frac{\pi}{2} \), yielding

\[
Z_{2l}(E, F) \leq \frac{2\rho \tilde{\rho}}{1 - 2\rho} \approx \frac{2\tilde{\rho}}{1 - 2\tilde{\rho}}, \quad \tilde{\rho} := \exp \left( -\frac{\pi^2}{2 \log \left( \frac{16}{1 - \delta^4} \right)} \right).
\]

Now, we prove that some matrices involved in the decomposition of the off-diagonal submatrices of \( \psi^{(h)}(z) \) enjoy a small displacement rank.

Proposition 4.5. Under the assumptions and the notation of Lemma 3.2 we have

\[
\tilde{C}^{(h)}(z^{2h}) = [I \ I] \cdot X^{(h)}(z)Y^{(h)}(z)^t \cdot [I \ I]^t \cdot [I \ 0]^t
\]

where

\[
X^{(h)}(z) := \begin{bmatrix} \mathcal{K}M_{2h}(\hat{G}_D, \hat{a}) \\ z^{1-2h} \mathcal{K}M_{2h}(G_D, a)J \end{bmatrix}, \quad Y^{(h)}(z) := \begin{bmatrix} z^{2h-1} \mathcal{K}M_{2h}(R^t, b)J \\ \mathcal{K}M_{2h}(\hat{R}^t, \hat{b}) \end{bmatrix}.
\]

Moreover, we have the following displacement relations:

\[
\rho_{W_D, \Pi}(X^{(h)}) = 1, \quad \rho_{W, \Pi}(Y^{(h)}) = 1,
\]

where

\[
W_D := \begin{bmatrix} \hat{G}_D & 0 \\ 0 & G_D^t \end{bmatrix}, \quad W := \begin{bmatrix} (R^t)^t & 0 \\ 0 & \hat{R}^t \end{bmatrix}, \quad \Pi := \begin{bmatrix} 0 & 1 \\ 1 & \ddots \\ \ddots & \ddots \\ 1 & 0 \end{bmatrix},
\]

with the super-script \( \dagger \) indicates the Moore-Penrose pseudoinverse.

Proof. The first claim simply follows by expanding the expression for \( \tilde{C}^{(h)}(z^{2h}) \) and by comparing it with equation (9). Concerning the displacement equations, a direct computation shows that the matrices

\[
W_D X^{(h)}(z) - X^{(h)}(z) \Pi \quad \text{and} \quad WY^{(h)}(z) - Y^{(h)}(z) \Pi
\]

have only the last column possibly different from zero. \( \square \)

The above result allows us to give a bound to the singular values of \( \tilde{C}^{(h)}(z) \).
Theorem 4.6. Let \( \varphi(z) = z^{-1}A_1 + A_0 + zA_1 \) be an \( m \times m \) matrix Laurent polynomial such that the CR — given by \( [1] \) — can be carried out without no breakdown, the splitting property \( [2] \) is verified, and \( \varphi(z) \) has quasiseparable rank 1 for every \( z \in T \). Assume that the matrices \( R \) and \( R' \) which solve the matrix equations \( [3] \) are diagonalizable by means of the two eigenvector matrices \( V_R \) and \( V_{R'} \), respectively. Assume that \( G_D \) and \( R \) are invertible. Then \( \varphi(z) \in \mathcal{DQ}(f) \) where

\[
    f(l) := \gamma \cdot Z_l(E, T),
\]

with \( \gamma \) a multiple of \( \max\{\kappa(V_R), \kappa(V_{R'})\} \) and \( E \) contains the eigenvalues of \( \varphi(z) \).

Proof. Notice that a generic off-diagonal matrix \( \hat{C}^{(h)}(z) \) in \( \psi^{(h)}(z) \) can be seen as a submatrix of

\[
    [I \ I]X^{(h)}(z)Y^{(h)}(z)^T [I \ I].
\]

In view of Proposition 4.5 we know that \( Y^{(h)}(z) \) has displacement rank 1. The displacement relation for \( Y^{(h)}(z) \) involves the matrices \( W \) and \( \Pi \) whose entries correspond to those of \( \varphi(z) \) and to the roots of the unity of order \( 2^h \), respectively. Moreover, \( W \) is diagonalizable by means of \( V_W := \text{diag}(V_R^{-1}, V_{R'}^{-1}) \). Therefore, applying Corollary 4.3 we can write

\[
    \sigma_{1+i}(Y^{(h)}(z)) \leq Z_t(E, T) \cdot \|Y^{(h)}(z)\|_2 \cdot \kappa(V_W).
\]

Since \( W \) is block-diagonal we have \( \kappa(V_W) = \max\{\kappa(V_R), \kappa(V_{R'})\} \). In particular we can bound the singular values of \( \hat{C}^{(h)}(z) \) with the quantity

\[
    \sigma_{1+i}(\hat{C}^{(h)}(z)) \leq 2 \cdot Z_t(E, T) \cdot \|X^{(h)}(z)\|_2 \cdot \|Y^{(h)}(z)\|_2 \cdot \kappa'(V_W). 
\]

Defining \( \gamma := 2 \cdot \kappa(V_W) \cdot \max_{h \in \mathbb{N}, z \in T} \frac{2\|X^{(h)}(z)\|_2 \cdot \|Y^{(h)}(z)\|_2}{\|Y^{(h)}(z)\|_2} \) we get the thesis. \( \square \)

The constant \( \gamma \) in the previous theorem is an index of how much the factorization \( X^{(h)}(z)Y^{(h)}(z)^T \) is unbalanced. This limitation is not present in the following result which describes the asymptotic behavior as \( h \to \infty \). It is possible to show that the block diagonal terms in \( W^{(h)}(z) = X^{(h)}(z)Y^{(h)}(z)^T \) quickly decay to 0 in practice, making the following bounds numerically accurate after a few steps.

Theorem 4.7. Let \( W^{(h)}(z) = X^{(h)}(z)Y^{(h)}(z)^T \), where \( X^{(h)}(z) \) and \( Y^{(h)}(z) \) are the matrices defined in Proposition 4.5. Then \( \lim_{h \to \infty} W^{(h)}(z) = W^{(\infty)} \) has the following block partitioning

\[
    W^{(\infty)} = \left[ \begin{array}{cc} 0 & B_1 \\ B_2 & 0 \end{array} \right]
\]

where the diagonal blocks are square and the off-diagonal blocks are independent of \( z \). Moreover, we have \( \rho_{V_D, V(W^{(\infty)})} = 2 \), where

\[
    V_D := \left[ \begin{array}{cc} \hat{G}_D & 0 \\ 0 & G_D \end{array} \right] \quad \text{and} \quad V := \left[ \begin{array}{cc} R & 0 \\ 0 & \hat{R} \end{array} \right].
\]

If the matrices \( G_D, \hat{G}_D, R \) and \( \hat{R} \) are diagonalizable by means of \( V_G, \hat{V}_G, V_R \) and \( V_{\hat{R}} \), respectively, then, indicating with \( \hat{C} \) the off-diagonal block in \( H_{\varphi, 0} \) corresponding to \( \hat{C}(z) \) we have the following bounds to its singular values

\[
    \sigma_{1+i}(\hat{C}) \leq \gamma \cdot Z_l(E, F), \quad \gamma := 2 \cdot \max\{\kappa(G), \kappa(\hat{G})\} \cdot \max\{\kappa(V_R), \kappa(V_{\hat{R}})\} \cdot \|\hat{C}\|_2,
\]

where \( E \) contains the eigenvalues of \( \varphi(z) \) and \( D(z) \) inside the unit disc while \( F \) contains those outside.
Proof. From the definition of $X^{(h)}$ and $Y^{(h)}$ one has

$$W^{(h)}(z) = \begin{bmatrix}
    z^{2h-1} \mathcal{K} \mathcal{M}_{2h}(\tilde{G}_D, \tilde{a}) J(\mathcal{K} \mathcal{M}_{2h}(R^t, b))^t & \mathcal{K} \mathcal{M}_{2h}(\tilde{G}_D, \tilde{a}) (\mathcal{K} \mathcal{M}_{2h}(\tilde{R}^t, \tilde{b}))^t \\
    \mathcal{K} \mathcal{M}_{2h}(G_D, a) (\mathcal{K} \mathcal{M}_{2h}(R^t, b))^t & z^{1-2h} \mathcal{K} \mathcal{M}_{2h}(G_D, a) J(\mathcal{K} \mathcal{M}_{2h}(\tilde{R}^t, \tilde{b}))^t
\end{bmatrix}.$$ 

Since the spectral radii of the matrices $R$, $\tilde{R}$, $G_D$ and $\tilde{G}_D$ are less than 1, then the block diagonal entries of $W^{(h)}$ tend to zero as $h \to \infty$ and the two off-diagonal blocks have limits $B_1$ and $B_2$, respectively. More precisely

$$W^{(\infty)} = \begin{bmatrix}
    0 & B_1 \\
    B_2 & 0
\end{bmatrix}, \quad B_1 = \sum_{i \geq 0} \tilde{G}_D \tilde{a} \tilde{b}^i \tilde{R}^i, \quad B_2 = \sum_{i \geq 0} G_D a b^i R^i.$$

Thus, we have

$$\tilde{G}_D B_1 - B_2 \tilde{R}^t = \tilde{a} \tilde{b}^t \tilde{R}^t.$$

An analogous argument holds for $B_2$, and gives the rank-2 displacement. The matrix $\tilde{C}$ can be written as $\tilde{C} = [I \ I] \cdot W^{(\infty)} \cdot [I \ I]^t \cdot [I \ 0]$, which corresponds to an off-diagonal block of $\lim_{h \to \infty} \psi^{(h)}(z)$. Due to the recurrence relation $\psi^{(h+1)}(z) = \frac{1}{2} (\psi^{(h)}(z) + \psi^{(h)}(-z))$, this limit is equal to the central coefficient $H_{0,0}$ in the series expansion of $\psi(z)$. The thesis follows by applying Corollary 4.3.

5 Experimental validation of the results

This section is devoted to verify the previous results by means of numerical experiments. We do that by computing numerical estimates of the bound given in Theorem 4.7 together with the singular values of the off-diagonal blocks of $H_{0,0}$. The actual bounds are obtained by choosing a particular family of rational functions that suit the considered problem. We will see that, even if our choices are relatively simple, and not optimal, they already provide sharp decay bounds in practice.

As a first example, we consider instances of the problem coming from the framework of Markov chains i.e., the sum $A_{-1} + A_0 + I + A_1$ is sub-stochastic, that is, it has non-negative entries and the sum along each row is at most 1. In particular, the matrices $A_{-1}$, $I + A_0$ and $A_1$ have non negative entries and are scaled in order to satisfy the splitting assumption (2) (see also Section 4.3 in [9]).

We select dense $300 \times 300$-blocks generated at random and such that $\varphi(z)$ is of quasiseparable rank 1. For satisfying the latter hypothesis we impose that the strictly triangular parts of the blocks are the restrictions of dyads with the same left vectors.

We divide the resulting distribution of the eigenvalues in three cluster. One is contained in a neighborhood of 0, another is in the complement of the disc of radius 4 and finally we have two eigenvalues close to 1, $\lambda_1$ and $\lambda_2$, inside and outside the unit circle, respectively.

Motivated by this, we choose the sequence of rational function

$$r_t(z) := \frac{z - \lambda_1}{z - \lambda_2} z^{l-1},$$

for roughly estimating the Zolotarev problem. The results are shown in Figure 2.

As a second example, we consider the linear system arising from the discretization of a 2D
Figure 2: Singular value decay in $H_0$ and the bound given by Theorem 4.7.
Poisson equation, whose matrix is block tridiagonal with the following form:

\[
T = \begin{bmatrix}
\tilde{A} & \tilde{C} \\
B & A & C \\
& \ddots & \ddots & \ddots \\
B & A & C \\
& & B & \tilde{A}
\end{bmatrix},
\quad A = \begin{bmatrix}
4 & -1 \\
-1 & 4 & -1 \\
& \ddots & \ddots & \ddots \\
-1 & 4 & -1 \\
& & -1 & 4
\end{bmatrix},
\quad B = C = -I.
\]

The above system can be solved by means of the cyclic reduction. The eigenvalues of the associated \(\varphi(z)\) can be computed explicitly and one can easily check that they are real positive and provide a splitting \(t = 1 - 1/(n + 1) + O(1/(n + 1)^2)\). The matrices \(A, B\) and \(C\) are very special instances of 1-quasiseparable matrices, so we can state a refined version of Theorem 4.7 which gives a smaller displacement rank for the limit case.

**Proposition 5.1.** Let \(A, B\) and \(C\) as above, and \(W^{(\infty)}(z)\) as defined in Theorem 4.7. If \(\tilde{C}\) is one off-diagonal block of \(H_{\psi,0}\) then

\[
\sigma_{1+l}(\tilde{C}) \leq \gamma \cdot Z_l(E, F), \quad \gamma := 2 \cdot \|	ilde{C}\|_2.
\]

**Proof.** Due to the symmetry properties of the coefficients \(A, B\) and \(C\), and to the palindromicity of \(\phi(z)\) and \(D(z)\), we have

\[
G_D = \hat{G}_D, \quad R = \hat{R}, \quad a = \hat{a}, \quad b = \hat{b}.
\]

In this way, we find that the matrix \([I \quad I] W^{(\infty)}[I \quad I]\) satisfies a displacement relation of rank 1 with the same matrices of Theorem 4.7. Therefore, the bound on the singular values holds with \(l\) instead of \(2l\). Moreover, since the matrices \(G, \hat{G}, R\) and \(\hat{R}\) can be diagonalized by means of orthogonal matrices, the maximum of their spectral conditioning is 1.

In order to verify the bound for this example we have carried out CR until convergence on a 200 \(\times\) 200 example and we have plotted the singular values of an off-diagonal block of the computed \(H_0\). Then, we have estimated the bound coming from Proposition 5.1 using a rational function of this form:

\[
\mathcal{r}_l(z) := (z - \delta) \prod_{j=1}^{l-1} \frac{z - q_j}{z - p_j}.
\]

The points \(q_j\) and \(p_j\) are chosen with a greedy approach as the maximizer and minimizer of \(r_{l-1}(z)\) in the sets \(E\) and \(F\) respectively. The point \(\delta\) is the rightmost eigenvalue of \(\varphi(z)\) inside the unit disc. The bound is compared with the one coming from Theorem 4.4 and with the one from [9]. The results are reported in Figure 3.

In this case the bound from [9] is useless since the approach used there relies on a wide splitting of the eigenvalues of \(\varphi(z)\). It is also interesting to note that even if the bound of Theorem 4.4 is optimal for real intervals an ad-hoc choice for the approximant in a discrete set can deliver better results.

### 6 Some applications

In this section we show some applications of CR in the case of quasiseparable blocks which for notational simplicity we refer to as Quasiseparable Cyclic Reduction (QCR for short), and we present some numerical results.
Figure 3: Decay of the singular values in one of the off-diagonal blocks of $H_0$ in the Laurent expansion of $\psi(z)$, computed by means of the CR for the Poisson matrix. We have reported the actual decay and the bounds obtained by means of Proposition 5.1, the results in [9], and Theorem 4.4.

A first application concerns solving a block tridiagonal linear system of the kind $A_n x = b$ where $A_n = \text{trid}_n(B, A, C)$, the blocks $A, B, C$ are $m \times m$ matrices such that CR can be carried out with no breakdown, the right-hand side vector $b$ and the unknown vector $x$ are partitioned into $n$ blocks $b_i$ and $x_i$, respectively of size $m$. For simplicity, assume $n = 2^k - 1$ so that the description of CR is simpler, for more details in the general case we refer the reader to [10].

The system can be written in the form

$$
\begin{bmatrix}
A & C \\
B & A & C \\
\vdots & \ddots & \ddots & \vdots \\
\vdots & \ddots & \ddots & \ddots & \ddots \\
B & C & \cdots & A \\
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
\vdots \\
x_n \\
\end{bmatrix}
= \begin{bmatrix}
b_1 \\
b_2 \\
\vdots \\
b_n \\
\end{bmatrix}, \quad A, B, C \in \mathbb{R}^{m \times m}, \quad x_i, b_i \in \mathbb{R}^m.
$$

An odd-even permutation of block rows and columns yields

$$
\begin{bmatrix}
A & \cdots & C \\
A & \cdots & B \\
\vdots & \ddots & \ddots & \ddots \\
\vdots & \ddots & \ddots & \ddots & \ddots \\
B & C & \cdots & A \\
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
\vdots \\
x_n \\
\end{bmatrix}
= \begin{bmatrix}
b_1 \\
b_2 \\
\vdots \\
b_n \\
\end{bmatrix}.
$$
Then one step of block Gaussian elimination performed to vanish the south-western block, yields

\[
\begin{bmatrix}
A & C \\
B & A
\end{bmatrix}
\begin{bmatrix}
A^{(1)} & C^{(1)} \\
B^{(1)} & A^{(1)}
\end{bmatrix}
\begin{bmatrix}
x_1 \\
\vdots \\
x_n
\end{bmatrix}
= 
\begin{bmatrix}
b_1 \\
\vdots \\
b_n
\end{bmatrix}
\]

with

\[
A^{(1)} = A - BA^{-1}C - CA^{-1}B, \\
B^{(1)} = -BA^{-1}B, \\
C^{(1)} = -CA^{-1}C,
\]

and

\[
b^{(1)}_i = b_{2i} - BA^{-1}b_{2i-1} - CA^{-1}b_{2i+1}, \quad i = 1, \ldots, n - 1.
\] (11)

The south-eastern block yields the system of the kind

\[
A_{n-\frac{1}{2}} \begin{bmatrix}
x_{even}
\end{bmatrix} = b^{(1)}
\]

with

\[
A_{n-\frac{1}{2}} = \text{trid}_{n-\frac{1}{2}}(B^{(1)}, A^{(1)}, C^{(1)}),
\]

where \(x_{even}\) denotes the subvector of \(x\) formed with the even block components, whose solution can be obtained by cyclically applying CR. Once the even block components of the block vector \(x\) have been computed, they can be substituted in the first part of the linear equations so that the odd block components of \(x\) are recovered. The hierarchical quasiseparability of the block matrices makes each operation of low cost.

Thus, the first (as well as the generic) step of CR performs the following steps

(i) Given the \(m \times m\) matrices \(A, B, C\) compute the matrices \(A^{(1)}, B^{(1)}, C^{(1)}\).

(ii) Given the \(m\)-vectors \(b, i = 1, \ldots, n\), compute \(b^{(1)}_i, i = 1, \ldots, \frac{n-1}{2}\) by means of (11).

(iii) Recursively solve the system \(\text{trid}_{n-\frac{1}{2}} x_{even} = b^{(1)}\) by means of CR.

(iv) Compute the odd components of the solution by mean of back substitution:

\[
x_1 = A^{-1}_1(b_1 - C_1 x_2),
\]

\[
x_i = A^{-1}_i(b_i - B x_{i-1} - C x_{i+1}), \quad i = 3, 5, \ldots, n - 2,
\]

\[
x_{n} = A^{-1}_n(b_n - B x_{n-1}),
\]

In the case where the blocks \(A, B, C\) are quasidecomparable, say, they are tridiagonal, not necessarily Toeplitz as in [13], and relying on the \(H\)-matrix representation as in [9], in view of the preservation of the hierarchical structure of the blocks \(A^{(h)}, B^{(h)}, C^{(h)}\) shown in the previous sections, the cost of step (i) is \(O(m \log^2 m)\), while the costs of steps (ii) and (iv) is \(O(n m \log m)\). Therefore, indicating with \(T(m, n)\) the asymptotic computational complexity of the whole algorithm with \(n = 2^k - 1\), we have

\[
T(m, n) = T \left( m, \frac{n - 1}{2} \right) + O(m \log^2 m) + O(n m \log m).
\]
Since \( T(m,1) = O(m \log^2 m) \), we obtain \( T(m,n) = O(mn \log m) + O(m \log^2 m \log n) \). For \( m = n \) this yields \( T(n,n) = O(n^2 \log n) + O(n \log^3 n) \).

It is interesting to recall that if \( A_n \) is the discrete Laplacian where \( A = \text{tridiag}(-1,4,-1) \), \( B = C = -I \) then CR has a cost of \( O(n^2 \log n) \) ops \([25]\) while the fast Poisson solvers based on the combination of Fourier analysis and CR \([16]\) have a cost of \( O(n^2 \log \log n) \) ops. Our approach has a slightly higher cost but covers a wider range of cases including block tridiagonal block Toeplitz matrices with banded (not necessarily Toeplitz) blocks.

Observe that CR preserves slightly more general structures than the block tridiagonal block Toeplitz. In particular it is possible to handle the case where the first and last blocks in the main diagonal differ from the other blocks on the same diagonal \([10]\).

### 6.1 Solving certain generalized Sylvester equations

For an \( m \times n \) matrix \( X \) denote \( x = \text{vec}(X) \) the \( mn \)-vector obtained by stacking the columns of \( X \). Then, for any pair of matrices \( A, B \) of compatible sizes, one has \( \text{vec}(AB) = (I \otimes A)\text{vec}(B) = (B^t \otimes I)\text{vec}(A) \).

Consider the linear matrix equation

\[
\sum_{i=1}^{s} A_i X B_i = C, \quad A_i \in \mathbb{R}^{m \times m}, \quad B_i \in \mathbb{R}^{n \times n}, \quad X, C \in \mathbb{R}^{m \times n},
\]

and suppose that \( B_i, i = 1, \ldots, s \) are tridiagonal Toeplitz matrices.

Applying the vec operator on both sides of (12) we get the \( mn \times mn \) linear system

\[
Wx = c, \quad W = \sum_{i=1}^{s} B_i^t \otimes A_i, \quad x = \text{vec}(X), \quad c = \text{vec}(C).
\]

Since each term \( B_i^t \otimes A_i \) is block tridiagonal and block Toeplitz, then the coefficient matrix of (13) is block tridiagonal, block Toeplitz as well. If the matrices \( A_i \) are \( k_i \)-quasiseparable then the blocks of \( W \) are \( k \)-quasiseparable with \( k = \sum_{i=1}^{s} k_i \). If \( k \) is negligible with respect to \( m \) then we may solve the generalized Sylvester equation by means of QCR.

### 6.2 Numerical results

A possible application of this algorithm is solving discretized partial differential equations coming from convection diffusion problems of the form

\[
-\epsilon \Delta u(x,y) + w \cdot \nabla u(x,y) = f(x,y), \quad \Omega \subset \mathbb{R}^2
\]

where \( u(x,y) \) is the unknown function, and we assume that the convection vector \( w \) depends only on one of the two coordinates. For simplicity we assume that it only depends on \( x \). According to \([25]\) we can discretize the above problem obtaining the following Sylvester equation in the matrix unknown \( U \):

\[
\epsilon T_1 U + \epsilon U T_2 + \Phi_1 B_1 U + \Phi_2 U B_2 = F.
\]

The independence on \( y \) of the convection vector ensures that all the right factors in the previous equation are almost Toeplitz. The matrices \( \Phi_i \) are diagonal while \( T_i \) and \( B_i \) arise from the discretization of the differential operators and they are all tridiagonal and Toeplitz with the exceptions of the first and last rows (due to the boundary conditions). The matrix \( F \) contains the evaluations of the function \( f \) on the discretized grid. We refer to \([25]\) for an in depth analysis.
Figure 4: Timings of the quasiseparable cyclic reduction (QCR) and the Sylvester solver implemented in the \texttt{lyap} function in MATLAB.

| Size | \(T_{QCR}\) (s) | \(Res_{QCR}\) | \(T_{lyap}\) (s) | \(Res_{lyap}\) |
|------|----------------|----------------|----------------|----------------|
| 100  | 0.12           | 2.16 \(\cdot\) \(10^{-13}\) | 1.83 \(\cdot\) \(10^{-2}\) | 1.18 \(\cdot\) \(10^{-12}\) |
| 200  | 0.27           | 1.54 \(\cdot\) \(10^{-12}\) | 4.99 \(\cdot\) \(10^{-2}\) | 5.56 \(\cdot\) \(10^{-12}\) |
| 400  | 0.85           | 5.53 \(\cdot\) \(10^{-12}\) | 0.29           | 5.17 \(\cdot\) \(10^{-11}\) |
| 800  | 3.2            | 4.19 \(\cdot\) \(10^{-11}\) | 2.06           | 9.04 \(\cdot\) \(10^{-11}\) |
| 1,600| 9.42           | 1.25 \(\cdot\) \(10^{-10}\) | 16.63          | 5.64 \(\cdot\) \(10^{-10}\) |
| 3,200| 23.86          | 6.78 \(\cdot\) \(10^{-10}\) | 142.78         | 2.06 \(\cdot\) \(10^{-9}\)  |
| 6,400| 58.79          | 2.41 \(\cdot\) \(10^{-9}\)  | 1.612          | 2.98 \(\cdot\) \(10^{-8}\)  |
| 12,800| 219.27        | 7.8 \(\cdot\) \(10^{-9}\)   | —              | —               |

We performed some numerical tests on one of the example in [25] namely (14) with \(\epsilon = 0.0333\) and \(w = (1 + (x+1)^2, 0)\). The right-hand side \(F\) is chosen at random. Since in this case \(\Phi_2 = 0\) the problem is reduced to solving the Sylvester equation

\[
(\epsilon T_1 + \Phi_1 B_1)U + U\epsilon T_2 = F.
\]

In Figure 4 and Table 1 we compare the timings and the residue of QCR with those of the function \texttt{lyap} from the control toolbox of MATLAB R2013a. Note that our approach can be applied even if the second coordinate of \(w\) is non zero and dependent only on \(x\). In fact, in this way we retrieve a generalized Sylvester equation that can be solved with this algorithm.

7 Concluding remarks

In this work we have provided an alternative analysis, with respect to [9], of the numerical preservation of the quasiseparable structures of the matrices generated by the cyclic reduction.
The theoretical results that we have obtained better describe the phenomenon in many instances coming from the applications. Examples related to the solution of Sylvester equations arising in the discretization of elliptic PDEs, and from queuing theory, have been shown.

The connection between the numerical preservation of the structure and the existence of accurate solutions of certain discrete rational approximation problems have been pointed out.

In the second part, the use of CR, together with hierarchical representations, as a direct method for the solution of block tridiagonal “almost” Toeplitz systems has been explored and has lead to the algorithm QCR. This procedure has an asymptotic complexity of \(O(mn \log m) + O(m \log^2 m \log n)\), where \(n\) is the number of the blocks and \(m\) their size. Applications to the solution of elliptic differential equations have been shown and the effectiveness of the approach reported.

Applications to solving certain generalized Sylvester equations, of the form

\[
\sum_{i=1}^{k} A_i XB_i = D,
\]

have been shown in the case where all the blocks \(B_i\)s are tridiagonal Toeplitz (possibly with only the first and last row with different entries), and the \(A_i\)s have a low quasiseparable rank. Under these hypothesis, and the assumption that the sum of the quasiseparable ranks of the \(A_i\)s is negligible compared to \(m\), the complexity of the method is \(O(m^2 \log m)\).

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