# A small note on the scaling of symmetric positive definite semiseparable matrices

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#### Abstract

In this paper we will adapt a known method for diagonal scaling of symmetric positive definite tridiagonal matrices towards the semiseparable case. Based on the fact that a symmetric, positive definite tridiagonal matrix T satisfies property A, one can easily construct a diagonal matrix (D) over cap such that (D) over capT (D) over cap has the lowest condition number over all matrices DTD, for any choice of diagonal matrix D. Knowing that semiseparable matrices are the inverses of tridiagonal matrices, one can derive similar properties for semiseparable matrices. Here, we will construct the optimal diagonal scaling of a semiseparable matrix, based on a new inversion formula for semiseparable matrices. Some numerical experiments are performed. In a first experiment we compare the condition numbers of the semiseparable matrices before and after the scaling. In a second numerical experiment we compare the scalability of matrices coming from the reduction to semiseparable form and matrices coming from the reduction to tridiagonal form.

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# A small note on the scaling of symmetric positive definite semiseparable matrices<sup>\*†</sup>

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#### Abstract

In this paper we will adapt a known method for diagonal scaling of symmetric positive definite tridiagonal matrices towards the semiseparable case. Based on the fact that a symmetric, positive definite tridiagonal matrix T satisfies property A, one can easily construct a diagonal matrix  $\hat{D}$  such that  $\hat{D}T\hat{D}$  has the lowest condition number over all matrices DTD, for any choice of diagonal matrix D. Knowing that semiseparable matrices are the inverses of tridiagonal matrices, one can derive similar properties for semiseparable matrices. Here, we will construct the optimal diagonal scaling of a semiseparable matrix, based on a new inversion formula for semiseparable matrices.

Some numerical experiments are performed. In a first experiment we compare the condition numbers of the semiseparable matrices before and after the scaling. In a second numerical experiment we compare the scalability of matrices coming from the reduction to semiseparable form and matrices coming from the reduction to tridiagonal form.

Keywords: Condition number, semiseparable matrices, property A, inversion of semiseparable matrices

AMS Classification: 65F35, 15A12

## 1 Introduction

The matrices we consider in this paper are symmetric and positive definite. A symmetric matrix A is said to have property A when there exists a permutation P, such that  $PAP^T$  is of the following form:

$$\hat{A} = PAP^{T} = \begin{pmatrix} D_{1} & F \\ F^{T} & D_{2} \end{pmatrix}, \tag{1}$$

where *F* is an arbitrary matrix and  $D_1$  and  $D_2$  are diagonal matrices. As the matrix is symmetric and positive definite the elements of  $D_1$  and  $D_2$  are positive. These matrices are known to be scaled best if the diagonal matrices  $D_1$  and  $D_2$  are equal to the identity matrix (see [2, 1, 5]). This can be achieved rather easily by applying a diagonal scaling of the matrix. Defining the scaling matrix  $\hat{D}$  as

$$\hat{D} = \begin{pmatrix} \frac{1}{\sqrt{(D_1)_{i,i}}} & \\ & \frac{1}{\sqrt{(D_2)_{i,i}}} \end{pmatrix}$$

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gives us the desired solution. In this paper we will use this result for tridiagonal matrices to derive the best diagonal scaling for semiseparable matrices. A tridiagonal matrix satisfies property A. For the permutation  $P^T$  one uses a reshuffling of the columns, thereby first placing the even columns ([2,4,...]) and after the even, the odd columns ([1,3,5,...]). Applying the same reshuffling technique P for the rows, one gets a matrix  $PTP^T$  of the desired form.

The optimal scaling matrix  $\hat{D}$  for semiseparable matrices is calculated by using an order O(n) inversion formula for semiseparable matrices. The most early paper related to the inversion of semiseparable and tridiagonal matrices is [3], in which irreducible symmetric tridiagonal matrices are inverted. The most early papers related to the inversion of semiseparable matrices are [6, 7]. These papers focus on semiseparable matrices arising in statistical applications. In fact inverting these matrices, as well as scaling these matrices is a very important subject in statistics. The scaling is related to the variance in weighted least squares regression coefficients [4] (in this paper attention is paid to the condition number of a specific semiseparable matrix, arising in statistical applications).

The present paper is organized as follows: in Section 2, we derive the structure of the optimal diagonal scaling matrix for semiseparable matrices. In Section 3 we give a method for computing the inverse of a semiseparable matrix. This method can now be used for calculating the optimal scaling matrix. In Section 4 we present numerical experiments showing the impact of the scaling on the condition number and on the accuracy when solving linear systems with semiseparable matrices as coefficient matrices.

#### 2 The scaling technique

Using the knowledge related to matrices having property A as described in the introduction, the optimal diagonal scaling of a tridiagonal matrix T can be obtained by choosing the diagonal  $\hat{D}$  in such a way that the diagonal elements of the tridiagonal matrix  $\hat{D}T\hat{D}$  are equal to one. This means that for a tridiagonal matrix T with diagonal elements  $a_i$ , our optimal scaling diagonal matrix  $\hat{D}$  has as diagonal elements  $1/\sqrt{a_i}$ . This is well defined as a symmetric positive definite matrix has positive diagonal elements.

Defining the condition number  $\kappa$  of a matrix A as  $\kappa(A) = ||A|| ||A^{-1}||$  for any consistent matrix norm, we have that  $\kappa(A) = \kappa(A^{-1})$ . This means that for an invertible symmetric, positive definite tridiagonal matrix T, with inverse S and optimal diagonal scaling matrix  $\hat{D}$  for T, we have:

$$\kappa(\hat{D}^{-1}S\hat{D}^{-1}) = \kappa(\hat{D}T\hat{D}) \le \kappa(DTD) = \kappa(D^{-1}SD^{-1}),$$

for any matrix *D*. Using the inversion formula for semiseparable matrices as presented in Section 3 we can easily calculate in O(n) flops the optimal diagonal scaling of a semiseparable matrix.

#### **3** Inversion of semiseparable matrices

Traditional inversion methods of semiseparable matrices are based on the generator representation. This means that the lower triangular part of a semiseparable matrix, as well as the upper triangular part are coming from rank 1 matrices. Based on the Givens-vector representation (see [9]), we will derive here an inversion formula for the class of semiseparable matrices representable in this way. Moreover, this class is more general than the class of generator representable matrices (see [9]). The algorithm will be designed for a  $4 \times 4$  semiseparable matrix, as this illustrates the general case. Our semiseparable matrix *S* is of the following form (see [9]):

$$S = \begin{pmatrix} c_1d_1 & c_2s_1d_1 & c_3s_2s_1d_1 & s_3s_2s_1d_1 \\ c_2s_1d_1 & c_2d_2 & c_3s_2d_2 & s_3s_2d_2 \\ c_3s_2s_1d_1 & c_3s_2d_2 & c_3d_3 & s_3d_3 \\ s_3s_2s_1d_1 & s_3s_2d_2 & s_3d_3 & d_4 \end{pmatrix}.$$

This matrix is stored by using 3n - 2 parameters namely the  $c_i, s_i$  and the  $d_i$  ( $d_1 \neq 0$  as the matrix is invertible). Note however that the  $c_i$  and  $s_i$  represent the cosine and sine of the same angle, this means that

essentially 2n - 1 parameters are needed to represent this matrix. (More information can be found in [9]). The inverse of the semiseparable matrix *S* is denoted as *T*:

$$T = \begin{pmatrix} a_1 & b_1 & & \\ b_1 & a_2 & b_2 & \\ & b_2 & a_3 & b_3 \\ & & & b_3 & a_4 \end{pmatrix}.$$

Let us start by calculating  $a_1$  and  $b_1$ . The 4 equations corresponding to the first column of the product ST = I, namely  $STe_1 = e_1$  (the elements  $e_i$  represent the standard basis vectors), are:

$$a_1(c_1d_1) + b_1(c_2s_1d_1) = 1, (2)$$

whereas, the remaining equations collapse into one singular equation: (This happens because the two first columns of S are dependent below the first row.)

$$a_1(s_1d_1) + b_1d_2 = 0. (3)$$

Rewriting equation (3) towards  $a_1d_1$  we get

$$a_1d_1=\frac{-b_1d_2}{s_1}.$$

If  $s_1 = 0$  we can easily see that  $b_1 = 0$  and  $a_1 = 1/d_1$ , so let us assume  $s_1$  to be different from zero.(We note that by definition of the representation the cosines are always positive. If  $s_1 = 0$  this implies  $c_1 = 1$ .) Substituting  $a_1d_1$  in equation (2) and rewriting the equation towards  $b_1$  gives us

$$b_1 = \frac{1}{c_2 s_1 d_1 - d_2 \frac{c_1}{s_1}}.$$

Note that this equation is well defined as  $s_1$  is different from zero and an easy calculation reveals that also the denominator in the equation has to be different from zero, otherwise the semiseparable matrix would be singular. Once we have  $b_1$  the calculation of  $a_1$  is straightforward using equation (3). We have calculated now  $a_1$  and  $b_1$ . To continue we calculate  $a_2$  and  $b_2$ . We write down the equations corresponding to the second column of ST = I, i.e.  $STe_2 = e_2$ , let us consider the equations on and below the diagonal. This gives us:

$$e_2^T ST e_2 = b_1(c_2 s_1 d_1) + a_2(c_2 d_2) + b_2(c_3 s_2 d_2) = 1,$$
(4)

while the equations  $e_3^T ST e_2 = 0$  and  $e_4^T ST e_2 = 0$  collapse again into one single equation:

$$b_1(s_2s_1d_1) + a_2(s_2d_2) + b_2d_3 = 0.$$
(5)

We will distinguish between two cases now:  $s_2 = 0$  and  $s_2 \neq 0$ .

• Let us assume  $s_2 = 0$ . This implies directly that  $b_2 = 0$ . Again we have to distinguish between two cases to calculate  $a_2$ . Assuming  $s_1 = 0$  we get  $a_2 = 1/d_2$ .

Assuming now  $s_1 \neq 0$ , we can write down equation  $e_1^T ST e_2 = 0$ :

$$b_1(c_1d_1) + a_2(c_2s_1d_1) = 0.$$

Rewriting this equation towards  $a_2$  we get  $a_2 = -b_1c_1/(c_2s_1)$ . This means that for  $s_2 = 0$  we can calculate  $a_2$  and  $b_2$ .

• Let us assume now  $s_2 \neq 0$  and  $f_2 = b_1 s_1 d_1$ , then we get for equations (4) and (5):

$$\begin{cases} c_2 f_2 + a_2(c_2 d_2) + b_2(c_3 s_2 d_2) = 1 \\ s_2 f_2 + a_2(s_2 d_2) + b_2 d_3 = 0. \end{cases}$$
(6)

Extracting  $a_2d_2$  from the second equation of (6) gives us:

$$a_2d_2 = \frac{-b_2d_3 - s_2f_2}{s_2}$$

(which is well defined as  $s_2 \neq 0$ ) filling it in the first equation of (6), we get:

$$b_2 = \frac{1}{c_3 s_2 d_2 - d_3 \frac{c_2}{s_2}},$$

which is similar to equation (3). The calculation of  $a_2$  is again a little more complicated. Two cases can occur:

- If  $d_2 = 0$  we use the equation  $e_1STe_2 = 0$  and extract  $a_2$  from this equation, this gives us:

$$a_2 = \frac{-b_2c_3s_2s_1 - b_1c_1}{c_2s_1}.$$

We know that  $c_2s_1$  has to be different from zero, otherwise the semiseparable matrix would have been singular.

- If  $d_2 \neq 0$  we use equation (5). Giving us

$$a_2 = \frac{-b_2d_3 - s_2f_2}{s_2d_2}$$

which is well defined because both  $s_2$  and  $d_2$  are different from zero.

This last procedure can similarly be repeated for the third column, leading to the same formula's for calculating  $a_3$  and  $b_3$ . Only the last column needs some extra attention. For the last column we consider the equations  $e_4^T ST e_4 = 1$  and  $e_3^T ST e_4 = 0$ :

$$\begin{cases} b_3(s_3d_3) + a_4d_4 &= 1\\ b_3(c_3d_3) + a_4(s_3d_3) &= 0 \end{cases}$$

Using these formulas we get

$$a_4 = \frac{1 - b_3 s_3 d_3}{d_4},$$

if  $d_4 \neq 0$  or else if  $d_4 = 0$ 

$$a_4 = -\frac{b_3c_3}{s_3},$$

because if  $d_4 = 0$ ,  $s_3$  has to be different from zero, otherwise the matrix would have been singular.

Even though we calculated only the inverse of a  $4 \times 4$  matrix, the procedure clearly demonstrates how to calculate inverses of larger semiseparable matrices. Moreover, inverting semiseparable matrices in this way leads to an O(n) procedure.

# 4 Numerical experiments

The software used for performing these numerical experiments can be downloaded from: *http://www.cs.kuleuven.ac.be/~marc*.

The experiments were performed using Matlab 7 Release 14.<sup>1</sup>.

For the first set of experiments we generated matrices of sizes ranging from 200 to 600, thereby varying the condition number from  $10^8$  to  $10^{14}$ . The table below shows the difference in condition number, with and without the diagonal scaling.

<sup>&</sup>lt;sup>1</sup>Matlab is a registered trademark of The Mathworks, inc.

$\kappa \approx$	size $\rightarrow$	200	300	400	500	600
108	without scaling	1.775e+08	9.197e+07	7.837e+07	7.803e+07	7.720e+07
	with scaling	2.783e+06	1.515e+07	7.528e+06	1.449e+07	7.874e+06
10 <sup>10</sup>	without scaling	1.319e+10	1.151e+10	1.012e+10	7.463e+09	6.387e+09
	with scaling	3.069e+08	1.325e+09	4.186e+08	5.071e+08	2.344e+08
10 <sup>12</sup>	without scaling	1.344e+12	1.432e+12	9.865e+11	1.065e+12	8.168e+11
	with scaling	1.846e+10	7.043e+09	5.135e+10	2.337e+10	4.045e+10
10 <sup>14</sup>	without scaling	1.259e+14	1.673e+14	6.909e+13	1.079e+14	6.556e+13
	with scaling	2.466e+11	2.209e+12	1.942e+12	9.188e+12	6.571e+11

Table 1: Scalability of semiseparable matrices.

Recently an orthogonal similarity reduction to semiseparable form was proposed [8]. This reduction performs  $9n^2$  more operations than the corresponding reduction to tridiagonal form. Both reductions have an overall cost of  $O(n^3)$ . These  $9n^2$  operations create an extra convergence behavior (nested subspace iteration) inside the reduction algorithm. This behavior is not present in the reduction to tridiagonal form. In a second experiment we compared the scalability of the matrices reduced to either tridiagonal or semiseparable form. Positive definite matrices of various condition numbers were generated and afterwards reduced to semiseparable and tridiagonal form. For both forms the scalability was tested. The results show the condition numbers of the matrices after the reduction without scaling, and then after scaling was applied. In the table with *S without* we denote the semiseparable matrix without scaling, with *T without*, the tridiagonal without scaling, and with *S with* or *T with*, these matrices with scaling.

$\kappa \approx$	size $\rightarrow$	200	300	400	500	600
108	S without	1.608e+08	6.012e+07	9.702e+07	7.306e+07	5.759e+07
	S with	2.044e+00	2.156e+00	2.046e+00	2.214e+00	2.017e+00
	T without	1.608e+08	6.012e+07	9.702e+07	7.306e+07	5.759e+07
	T with	1.070e+06	3.940e+06	1.096e+07	4.363e+06	6.862e+06
10 <sup>10</sup>	S without	1.482e+10	7.719e+09	7.892e+09	1.261e+10	7.448e+09
	S with	1.893e+00	2.190e+00	1.954e+00	1.970e+00	1.890e+00
	T without	1.482e+10	7.719e+09	7.892e+09	1.261e+10	7.448e+09
	T with	1.397e+08	1.057e+09	8.510e+08	3.860e+08	1.984e+08
10 <sup>12</sup>	S without	1.637e+12	1.071e+12	9.664e+11	7.772e+11	7.576e+11
	S with	2.160e+00	1.905e+00	1.809e+00	2.115e+00	1.898e+00
	T without	1.637e+12	1.071e+12	9.664e+11	7.772e+11	7.576e+11
	T with	7.110e+10	2.697e+10	5.145e+10	1.937e+10	1.213e+10
10 <sup>14</sup>	S without	1.344e+14	1.394e+14	1.617e+14	8.475e+13	8.471e+13
	S with	2.091e+00	2.045e+00	1.842e+00	1.809e+00	2.238e+00
	T without	1.637e+12	1.071e+12	9.664e+11	7.772e+11	7.576e+11
	T with	1.138e+12	2.115e+12	5.823e+11	2.656e+12	3.105e+12

Table 2: Comparison in scalability.

This indicates that due to extra convergence properties of the reduction to semiseparable form [8], these matrices are far better scalable than the matrices coming from the reduction to tridiagonal form.

## 5 Conclusions and future research

In this paper an O(n) scaling technique for semiseparable matrices was presented, based on the scaling of tridiagonal matrices.

Future research will focus on a direct translation of property A towards the semiseparable case, as this could lead to scaling of more general classes of semiseparable matrices, such as sequentially or hierarchically semiseparable ones. Also attention will be paid towards scaling techniques for rectangular matrices, as this can improve accuracy e.g. in least squares problems.

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