

A PARAMETERIZED LOWER BOUND FOR THE SMALLEST SINGULAR VALUE*

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Abstract. This paper presents a parameterized lower bound for the smallest singular value of a matrix based on a new Geršgorin-type inclusion region that has been established recently by these authors. The comparison of the new lower bound with known ones is supplemented with a numerical example.

Key words. Eigenvalue, Inclusion region, Smallest singular value, Lower bound.

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1. Introduction. To estimate matrix singular values is an attractive topic in matrix theory and numerical analysis, especially to give a lower bound for the smallest one. Let $A = (a_{ij}) \in \mathbb{C}^{n \times n}$ and let A^* be the conjugate transpose of A . Then the singular values of A are the square roots of the eigenvalues of AA^* . Throughout the paper we use $\sigma_n(A)$ to denote the smallest singular value of A . Denote $N := \{1, 2, \dots, n\}$. Define, for all $k \in N$,

$$r_k(A) := \sum_{j \in N \setminus \{k\}} |a_{kj}|, \quad c_k(A) := \sum_{j \in N \setminus \{k\}} |a_{jk}|, \quad h_k(A) := \frac{1}{2} (r_k(A) + c_k(A)).$$

In the past three decades, several useful lower bounds for the smallest singular value of a matrix have been presented in the literature; see Varah [7] and Qi [6]. By using Geršgorin's theorem (see Chapter 6 of [1]), Johnson [4] obtained a lower bound for $\sigma_n(A)$:

$$(1.1) \quad \sigma_n(A) \geq \min_{k \in N} \{|a_{kk}| - h_k(A)\}.$$

Recently, Johnson and Szulc [5] provided several further lower bounds for the

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smallest singular value. Two of them are

$$(1.2) \quad \sigma_n(A) \geq \frac{1}{2} \min_{k \in N} \left\{ \left(4|a_{kk}|^2 + [r_k(A) - c_k(A)]^2 \right)^{\frac{1}{2}} - 2h_k(A) \right\}$$

and

$$(1.3) \quad \sigma_n(A) \geq \frac{1}{2} \min_{i \neq k} \left\{ |a_{ii}| + |a_{kk}| - \left[(|a_{ii}| - |a_{kk}|)^2 + 4h_i(A)h_k(A) \right]^{\frac{1}{2}} \right\}.$$

In this paper, after introducing a new inclusion region for eigenvalues of a matrix in Section 2, we present a parameterized lower bound for the smallest singular value in Section 3. In Section 4, a numerical example is given to compare our results with the known ones.

2. Inclusion regions for eigenvalues. Recently, a new Geršgorin-type inclusion region for eigenvalues of a matrix has been provided by Huang, Zhang, and Shen in [3]. To introduce the result, we first present some notation. Let S be a nonempty subset of N . Denote $\bar{S} := N \setminus S$ and let $\mathcal{P}(N)$ denote the power set of N . For $A = (a_{ij}) \in \mathbb{C}^{n \times n}$, define, for all $i \in N$

$$r_i^S(A) := \sum_{k \in S \setminus \{i\}} |a_{ik}|, \quad r_i^{\bar{S}}(A) := \sum_{k \in \bar{S} \setminus \{i\}} |a_{ik}|,$$

$$c_i^S(A) := \sum_{k \in S \setminus \{i\}} |a_{ki}|, \quad c_i^{\bar{S}}(A) := \sum_{k \in \bar{S} \setminus \{i\}} |a_{ki}|.$$

If S contains a single element, say $S = \{i_0\}$, then we let $r_{i_0}^S(A) = 0$. Similarly $r_{i_0}^{\bar{S}}(A) = 0$ if $\bar{S} = \{i_0\}$. We sometimes use r_i^S (c_i^S , $r_i^{\bar{S}}$, $c_i^{\bar{S}}$) to denote $r_i^S(A)$ ($c_i^S(A)$, $r_i^{\bar{S}}(A)$, $c_i^{\bar{S}}(A)$, respectively). Define, for all $i \in S$ and $j \in \bar{S}$,

$$G_i^S(A) := \{z \in \mathbb{C} : |z - a_{ii}| \leq r_i^S\}, \quad G_j^{\bar{S}}(A) := \{z \in \mathbb{C} : |z - a_{jj}| \leq r_j^{\bar{S}}\}$$

and

$$G_{i,j}^S(A) := \left\{ z \in \mathbb{C} : z \notin G_i^S(A) \cup G_j^{\bar{S}}(A), (|z - a_{ii}| - r_i^S) (|z - a_{jj}| - r_j^{\bar{S}}) \leq r_i^{\bar{S}} r_j^S \right\}.$$

PROPOSITION 2.1. ([3]) *Let $A = (a_{ij}) \in \mathbb{C}^{n \times n}$. Then all the eigenvalues of A are located in*

$$G^S(A) := \left(\bigcup_{i \in S} G_i^S(A) \right) \cup \left(\bigcup_{j \in \bar{S}} G_j^{\bar{S}}(A) \right) \cup \left(\bigcup_{i \in S, j \in \bar{S}} G_{i,j}^S(A) \right).$$

This result is interesting, but it can not be applied directly to estimate $\sigma_n(A)$. So we must deduce other inclusion regions. Let $\alpha \in [0, 1]$ be given. For $i \in S$, $j \in \bar{S}$, define the following regions in the complex plane:

$$U_{i,j}^S(A) := \left\{ z \in \mathbb{C} : |z - a_{ii}| - r_i^S \leq \left(r_i^{\bar{S}} r_j^S \right)^\alpha \right\},$$

$$V_{i,j}^S(A) := \left\{ z \in \mathbb{C} : |z - a_{jj}| - r_j^{\bar{S}} \leq \left(r_i^{\bar{S}} r_j^S \right)^{1-\alpha} \right\},$$

PROPOSITION 2.2. Let $A = (a_{ij}) \in \mathbb{C}^{n \times n}$. Then all the eigenvalues of A are located in

$$K^S(A) := \left(\bigcup_{i \in S, j \in \bar{S}} U_{i,j}^S(A) \right) \cup \left(\bigcup_{i \in S, j \in \bar{S}} V_{i,j}^S(A) \right).$$

Proof. It is sufficient to show that $G^S(A) \subset K^S(A)$. Note that $G_i^S(A) \subseteq U_{i,j}^S(A)$ and $G_j^{\bar{S}}(A) \subseteq V_{i,j}^S(A)$. Therefore, for any $z \in G^S(A)$, if

$$z \in \bigcup_{i \in S} G_i^S(A) \quad \text{or} \quad z \in \bigcup_{j \in \bar{S}} G_j^{\bar{S}}(A),$$

then $z \in K^S(A)$. Otherwise, there exist $i_0 \in S$ and $j_0 \in \bar{S}$, such that

$$\left(|z - a_{i_0 i_0}| - r_{i_0}^S \right) \left(|z - a_{j_0 j_0}| - r_{j_0}^{\bar{S}} \right) \leq r_{i_0}^{\bar{S}} r_{j_0}^S = \left(r_{i_0}^{\bar{S}} r_{j_0}^S \right)^\alpha \left(r_{i_0}^{\bar{S}} r_{j_0}^S \right)^{1-\alpha}$$

which leads to

$$|z - a_{i_0 i_0}| - r_{i_0}^S \leq \left(r_{i_0}^{\bar{S}} r_{j_0}^S \right)^\alpha \quad \text{or} \quad |z - a_{j_0 j_0}| - r_{j_0}^{\bar{S}} \leq \left(r_{i_0}^{\bar{S}} r_{j_0}^S \right)^{1-\alpha}.$$

Hence

$$z \in U_{i_0, j_0}^S(A) \cup V_{i_0, j_0}^S(A).$$

And then $z \in K^S(A)$. $\square$

3. Main results. In this section we use the inclusions derived in Section 2 to estimate $\sigma_n(A)$. Denote the Hermitian part of A by

$$H(A) := \frac{1}{2}(A + A^*).$$

Let $\lambda_{\min}(H(A))$ be the smallest eigenvalue of $H(A)$. It is known that $\lambda_{\min}(H(A))$ is a lower bound for $\sigma_n(A)$ [2, p.227]. Moreover, define for all $i \in S$, $j \in \bar{S}$, and $\alpha \in [0, 1]$,

$$P_{i,j}^S(A) := |a_{ii}| - \frac{1}{2} [r_i^S(A) + c_i^S(A)] - \left[\frac{1}{4} (r_i^{\bar{S}}(A) + c_i^{\bar{S}}(A)) (r_j^S(A) + c_j^S(A)) \right]^\alpha,$$

$$Q_{i,j}^S(A) := |a_{jj}| - \frac{1}{2} [r_j^{\bar{S}}(A) + c_j^{\bar{S}}(A)] - \left[\frac{1}{4} (r_i^{\bar{S}}(A) + c_i^{\bar{S}}(A)) (r_j^S(A) + c_j^S(A)) \right]^{1-\alpha}.$$

THEOREM 3.1. Let $A = (a_{ij}) \in \mathbb{C}^{n \times n}$. Then

$$(3.1) \quad \sigma_n(A) \geq \min_{i \in S, j \in \bar{S}} \{P_{i,j}^S(A), Q_{i,j}^S(A)\}.$$

Proof. We first define a diagonal matrix $D = \text{diag}(e^{i\theta_1}, \dots, e^{i\theta_n})$, where $e^{i\theta_k} a_{kk} = |a_{kk}|$ if $a_{kk} \neq 0$ and $\theta_k = 0$ if $a_{kk} = 0$, $k \in N$. Since D is unitary, the singular values of DA are the same as those of A . Consequently, we have

$$(3.2) \quad \sigma_n(A) = \sigma_n(DA) \geq \lambda_{\min}(H(DA)).$$

Denote $B = (b_{kl}) := H(DA) = \frac{1}{2}(DA + A^*D^*)$. Thus, $b_{kk} = |a_{kk}|$, for $k \in N$, and

$$b_{kl} = \frac{1}{2} (e^{k\theta_k} a_{kl} + \bar{a}_{lk} e^{-k\theta_k}), \quad \text{for all } k \neq l, \quad k, l \in N.$$

Since B is a Hermitian matrix, its eigenvalues are all real. Let $\lambda_{\min}(B)$ denote the smallest eigenvalue of B . Then, by using Proposition 2.2, $\lambda_{\min}(B)$ must satisfy at least one of the following conditions

$$\lambda_{\min}(B) \geq |b_{ii}| - r_i^S(B) - [r_i^{\bar{S}}(B) r_j^S(B)]^\alpha, \quad i \in S, j \in \bar{S},$$

$$\lambda_{\min}(B) \geq |b_{jj}| - r_j^{\bar{S}}(B) - [r_i^{\bar{S}}(B) r_j^S(B)]^{1-\alpha}, \quad i \in S, j \in \bar{S}.$$

It follows that

$$\lambda_{\min}(B) \geq \min_{i \in S, j \in \bar{S}} \left\{ |b_{ii}| - r_i^S(B) - [r_i^{\bar{S}}(B) r_j^S(B)]^\alpha, \right. \\ \left. |b_{jj}| - r_j^{\bar{S}}(B) - [r_i^{\bar{S}}(B) r_j^S(B)]^{1-\alpha} \right\}.$$

By applying the triangle inequality, we have

$$\begin{aligned} & |b_{ii}| - r_i^S(B) - [r_i^{\bar{S}}(B) r_j^S(B)]^\alpha \\ &= |a_{ii}| - r_i^S(H(DA)) - [r_i^{\bar{S}}(H(DA)) r_j^S(H(DA))]^\alpha \\ &\geq |a_{ii}| - \frac{1}{2} (r_i^S(A) + c_i^S(A)) - \left[\frac{1}{4} (r_i^{\bar{S}}(A) + c_i^{\bar{S}}(A)) (r_j^S(A) + c_j^S(A)) \right]^\alpha \\ &= P_{i,j}^S(A). \end{aligned}$$

Similarly, one can obtain

$$|b_{ii}| - r_i^{\bar{S}}(B) - \left[r_i^{\bar{S}}(B) r_j^S(B) \right]^{1-\alpha} \geq Q_{i,j}^S(A).$$

Then from (3.2), we have

$$\sigma_n(A) \geq \min_{i \in S, j \in \bar{S}} \{P_{i,j}^S(A), Q_{i,j}^S(A)\}. \quad \square$$

Since the bound (3.1) holds for any nonempty $S \in \mathcal{P}(N)$ and any $\alpha \in [0, 1]$, we have obtain the following corollaries:

$$\text{COROLLARY 3.2. } \sigma_n(A) \geq \max_{S \in \mathcal{P}(N)} \max_{\alpha \in [0, 1]} \min_{i \in S, j \in \bar{S}} \{P_{i,j}^S(A), Q_{i,j}^S(A)\}.$$

$$\text{COROLLARY 3.3. ([4]) } \sigma_n(A) \geq \min_{i \in N} \left\{ |a_{ii}| - \frac{1}{2} (r_i(A) + c_i(A)) \right\}.$$

4. Numerical example. In this section, we give a numerical example to compare our bound (3.1) with known ones.

EXAMPLE 4.1. Consider the following matrices

$$A_1 = \begin{bmatrix} 11 & 5 & 6 \\ 4 & 12 & -5 \\ 3 & 4 & 13 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 18 & 2 & -5 \\ 6 & 15 & 8 \\ -6 & -3 & 17 \end{bmatrix}, \quad A_3 = \begin{bmatrix} 6 & 2 & -1 \\ 2 & 9 & 1 \\ 2 & -2 & -13 \end{bmatrix}.$$

Table 1. Comparison of lower bounds for $\sigma_n(A)$

Matrix	$\sigma_n(A_i)$	S	α	(1.1)	(1.2)	(1.3)	(3.1)
A_1	5.8446	$\{1\}$	0.57	2.0000	2.1803	2.4861	2.5886
A_2	9.6861	$\{2\}$	0.56	5.5000	6.1172	5.7827	5.7989
A_3	4.5433	$\{3\}$	0.10	2.5000	3.6921	2.3028	2.8377

From Table 1, we can see that bounds (1.2), (1.3), (3.1) are not comparable. Note that the tightness of our bounds depend on the choice of S and α in which, unfortunately, we do not find a method such that the derived bounds are optimal. However, these parameters offer the possibility to optimize the estimation. We hope that future research will propose a method to determine the parameters S and α that can give tighter lower bounds.

REFERENCES

- [1] R. A. Horn and C. R. Johnson. *Matrix Analysis*. Cambridge University Press, Cambridge, 1985.
- [2] R. A. Horn and C. R. Johnson. *Topics in Matrix Analysis*. Cambridge University Press, Cambridge, 1991.
- [3] T. Z. Huang, W. Zhang, and S. Q. Shen. Regions containing eigenvalues of a matrix. *Electron. J. Linear Algebra*, 15:215–224, 2006.
- [4] C. R. Johnson. A Gersgorin-type lower bound for the smallest singular value. *Linear Algebra Appl.*, 112:1–7, 1989.
- [5] C. R. Johnson and T. Szulc. Further lower bounds of the smallest singular value. *Linear Algebra Appl.*, 272:69–179, 1998.
- [6] L. Qi. Some simple estimates for singular values of a matrix. *Linear Algebra Appl.*, 56:105–119, 1984.
- [7] J. M. Varah. A lower bound for the smallest singular value. *Linear Algebra Appl.*, 11:3–5, 1975.