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# Bounds on Eigenvalues of Interval Matrices 

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Technical report No. 688

October 1996

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# Bounds on Eigenvalues of Interval Matrices ${ }^{1}$ 

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#### Abstract

We describe a rectangle in the complex plane enclosing all eigenvalues of an interval matrix $A^{I}$. We give theoretical bounds (Theorem 1) that are exact for symmetric or skew-symmetric matrices (Theorem 2) and practical bounds (Theorem 3) requiring evaluation of 6 minimal or maximal eigenvalues of symmetric matrices. Some consequences are mentioned.


## Keywords

Interval matrix, eigenvalue, bound

[^0]
## 1 Theoretical bounds

We consider square interval matrices in the form

$$
A^{I}=\left[A_{c}-\Delta, A_{c}+\Delta\right]=\left\{A ; A_{c}-\Delta \leq A \leq A_{c}+\Delta\right\}
$$

where inequalities are understood componentwise; thus $A_{c}$ is the center matrix and $\Delta$ is the radius matrix of $A^{I}$.

Theorem 1 Let $A^{I}=\left[A_{c}-\Delta, A_{c}+\Delta\right]$ be a square interval matrix. Then for each eigenvalue $\lambda$ of each $A \in A^{I}$ we have

$$
\begin{align*}
& \underline{r} \leq \operatorname{Re} \lambda \leq \bar{r},  \tag{1.1}\\
& \underline{i} \leq \operatorname{Im} \lambda \leq \bar{n} \tag{1.2}
\end{align*}
$$

where

$$
\begin{aligned}
\underline{r} & =\min _{\|x\|_{2}=1}\left(x^{T} A_{c} x-|x|^{T} \Delta|x|\right), \\
\bar{r} & =\max _{\|x\|_{2}=1}\left(x^{T} A_{c} x+|x|^{T} \Delta|x|\right), \\
\underline{i} & =\min _{\|\left(x_{1}, x_{2} \|_{2}=1\right.}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}-\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right), \\
\bar{\imath} & =\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right) .
\end{aligned}
$$

Comments. Vectors are always considered column vectors, so that $x^{T} y$ is the scalar product whereas $x y^{T}$ is the matrix $\left(x_{i} y_{j}\right)$. In the formulae for $\underline{i}$ and $\bar{\imath}$, for typographic reasons we write " $\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1$ " in the subscript instead of the correct $"\left\|\left(x_{1}^{T}, x_{2}^{T}\right)^{T}\right\|_{2}=1 "$. For $A=\left(a_{i j}\right)$ and $B=\left(b_{i j}\right)$ we use

$$
A \circ B=\sum_{i j} a_{i j} b_{i j}
$$

("scalar product of matrices"). Then we have

$$
x^{T} A y=\sum_{i j} x_{i} a_{i j} y_{j}=A \circ\left(x y^{T}\right) .
$$

Proof. Let $\lambda=\lambda_{1}+\lambda_{2} i$ be an eigenvalue of some $A \in A^{I}$. Then

$$
\begin{equation*}
A\left(x_{1}+x_{2} i\right)=\left(\lambda_{1}+\lambda_{2} i\right)\left(x_{1}+x_{2} i\right) \tag{1.3}
\end{equation*}
$$

for some real vectors $x_{1}, x_{2}, x_{1} \neq 0$ or $x_{2} \neq 0$, which may be normalized to achieve

$$
\begin{equation*}
x_{1}^{T} x_{1}+x_{2}^{T} x_{2}=1 . \tag{1.4}
\end{equation*}
$$

Premultiplying (1.3) by the complex conjugate vector $x_{1}-x_{2}$ i, we obtain

$$
\lambda_{1}+\lambda_{2} i=\left(x_{1}-x_{2} i\right)^{T} A\left(x_{1}+x_{2} i\right)
$$

which yields

$$
\begin{align*}
& \operatorname{Re} \lambda=\lambda_{1}=x_{1}^{T} A x_{1}+x_{2}^{T} A x_{2},  \tag{1.5}\\
& \operatorname{Im} \lambda=\lambda_{2}=x_{1}^{T} A x_{2}-x_{2}^{T} A x_{1} \text {. } \tag{1.6}
\end{align*}
$$

1) To prove that $\operatorname{Re} \lambda \leq \bar{r}$, denote $r(A)=\max _{\|x\|_{2}=1} x^{T} A x$, then we have

$$
\begin{aligned}
& x_{1}^{T} A x_{1} \leq r(A) x_{1}^{T} x_{1}, \\
& x_{2}^{T} A x_{2} \leq r(A) x_{2}^{T} x_{2},
\end{aligned}
$$

hence

$$
\begin{equation*}
x_{1}^{T} A x_{1}+x_{2}^{T} A x_{2} \leq r(A)\left(x_{1}^{T} x_{1}+x_{2}^{T} x_{2}\right)=r(A) \tag{1.7}
\end{equation*}
$$

due to (1.4), and

$$
\begin{align*}
r(A) & =\max _{\|x\|_{2}=1} x^{T} A x=\max _{\|x\|_{2}=1}\left(x^{T} A_{c} x+x^{T}\left(A-A_{c}\right) x\right)  \tag{1.8}\\
& \leq \max _{\|x\|_{2}=1}\left(x^{T} A_{c} x+|x|^{T} \Delta|x|\right)=\bar{r} .
\end{align*}
$$

Then from (1.5), (1.7) and (1.8) we obtain

$$
\operatorname{Re} \lambda \leq \bar{r}
$$

which is the right-hand side inequality in (1.1).
2) Since $-\lambda$ is an eigenvalue of $-A$ which belongs to $\left[-A_{c}-\Delta,-A_{c}+\Delta\right]$, from the result proved in 1) applied to $\left[-A_{c}-\Delta,-A_{c}+\Delta\right]$ we obtain

$$
-\operatorname{Re} \lambda=\operatorname{Re}(-\lambda) \leq \max _{\|x\|_{2}=1}\left(-x^{T} A_{c} x+|x|^{T} \Delta|x|\right)
$$

which implies

$$
\operatorname{Re} \lambda \geq-\max _{\|x\|_{2}=1}\left(-x^{T} A_{c} x+|x|^{T} \Delta|x|\right)=\min _{\|x\|_{2}=1}\left(x^{T} A_{c} x-|x|^{T} \Delta|x|\right)=\underline{r},
$$

which is the left-hand side inequality in (1.1).
3) Since

$$
\begin{aligned}
x_{1}^{T} A x_{2}-x_{2}^{T} A x_{1} & =x_{1}^{T} A_{c} x_{2}-x_{2}^{T} A_{c} x_{1}+x_{1}^{T}\left(A-A_{c}\right) x_{2}-x_{2}^{T}\left(A-A_{c}\right) x_{1} \\
& =x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\left(A-A_{c}\right) \circ\left(x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right) \\
& \leq x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|,
\end{aligned}
$$

from (1.6) and (1.4) we get

$$
\operatorname{Im} \lambda \leq \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right)=\bar{\imath},
$$

which is the right-hand side inequality in (1.2).
4) Since $-\lambda$ is an eigenvalue of $-A \in\left[-A_{c}-\Delta,-A_{c}+\Delta\right]$, applying the result in 3) we obtain

$$
-\operatorname{Im} \lambda=\operatorname{Im}(-\lambda) \leq \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}^{T}-A_{c}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right)
$$

and thereby also

$$
\operatorname{Im} \lambda \geq \min _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}-\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right)=\underline{i},
$$

which concludes the proof.

## 2 The bounds are exact in special cases

A real matrix $A$ is called symmetric if $A^{T}=A$ and skew-symmetric if $A^{T}=-A$. An interval matrix $A^{I}$ is said to be symmetric if

$$
A^{I T}=A^{I}
$$

and skew-symmetric if

$$
A^{I T}=-A^{I},
$$

where

$$
A^{I T}=\left\{A^{T} ; A \in A^{I}\right\}
$$

and

$$
-A^{I}=\left\{-A ; A \in A^{I}\right\} .
$$

Hence, $A^{I}=\left[A_{c}-\Delta, A_{c}+\Delta\right]$ is symmetric if and only if $\left[A_{c}^{T}-\Delta^{T}, A_{c}^{T}+\Delta^{T}\right]=$ $\left[A_{c}-\Delta, A_{c}+\Delta\right]$, which is equivalent to symmetry of both $A_{c}$ and $\Delta$. Similarly, $A^{I}$ is skew-symmetric if and only if $A_{c}$ is skew-symmetric and $\Delta$ is symmetric.

Theorem 2 The bounds (1.1) are exact (i.e., achieved over $A^{I}$ ) if $A^{I}$ is symmetric and the bounds (1.D) are exact if $A^{I}$ is skew-symmetric.

Proof. 1) Let $A^{I}$ be symmetric, so that $A_{c}$ and $\Delta$ are symmetric. Since the continuous mapping $x \mapsto x^{T} A_{c} x+|x|^{T} \Delta|x|$ achieves its maximum over the unit sphere $\left\{x ;\|x\|_{2}=1\right\}$, there exists an $x$ satisfying

$$
\begin{equation*}
\bar{r}=x^{T} A_{c} x+|x|^{T} \Delta|x| \tag{2.1}
\end{equation*}
$$

and $\|x\|_{2}=1$. Define a diagonal matrix $S$ by

$$
S_{j j}=\left\{\begin{aligned}
1 & \text { if } x_{j} \geq 0, \\
-1 & \text { if } x_{j}<0
\end{aligned}\right.
$$

$(j=1, \ldots, n)$, then $|x|=S x$ and from (2.1) we have

$$
\begin{equation*}
\bar{r}=x^{T} A_{c} x+x^{T} S \Delta S x=x^{T}\left(A_{c}+S \Delta S\right) x \leq \lambda_{\max }\left(A_{c}+S \Delta S\right), \tag{2.2}
\end{equation*}
$$

where $\lambda_{\max }\left(A_{c}+S \Delta S\right)$ denotes the maximal eigenvalue of $A_{c}+S \Delta S$ (which is symmetric since both $A_{c}$ and $\Delta$ are symmetric). Since $|S \Delta S|=\Delta$, the matrix $A_{c}+S \Delta S$ belongs to $A^{I}$, hence

$$
\lambda_{\max }\left(A_{c}+S \Delta S\right) \leq \bar{r}
$$

by Theorem 1, which combined with (2.2) gives

$$
\bar{r}=\lambda_{\max }\left(A_{c}+S \Delta S\right)
$$

hence $\bar{r}$ is achieved over $A^{I}$ (even more, it is achieved at a symmetric matrix in $A^{I}$, cf. Hertz [1]). The proof for $\underline{r}$ is analogous; in this case we obtain

$$
\underline{r}=\lambda_{\min }\left(A_{c}-S \Delta S\right)
$$

2) Let $A^{I}$ be skew-symmetric, so that $A_{c}$ is skew-symmetric and $\Delta$ is symmetric. We have

$$
\begin{equation*}
\bar{\imath}=x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right| \tag{2.3}
\end{equation*}
$$

for some $x_{1}, x_{2}$ satisfying $\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1$. Define

$$
z_{i j}=\left\{\begin{aligned}
-1 & \text { if }\left(x_{1}\right)_{i}\left(x_{2}\right)_{j}-\left(x_{2}\right)_{i}\left(x_{1}\right)_{j}<0 \\
0 & \text { if }\left(x_{1}\right)_{i}\left(x_{2}\right)_{j}-\left(x_{2}\right)_{i}\left(x_{1}\right)_{j}=0 \\
1 & \text { if }\left(x_{1}\right)_{i}\left(x_{2}\right)_{j}-\left(x_{2}\right)_{i}\left(x_{1}\right)_{j}>0
\end{aligned}\right.
$$

$(i, j=1, \ldots, n)$, then $z_{i j}=-z_{j i}$ for each $i, j$, hence the matrix $\tilde{\Delta}$ defined by

$$
\tilde{\Delta}_{i j}=z_{i j} \Delta_{i j}
$$

$(i, j=1, \ldots, n)$ is skew-symmetric (since $\Delta$ is symmetric). Let

$$
A=A_{c}+\tilde{\Delta}
$$

then $A \in A^{I}$ and $A$ is skew-symmetric (since both $A_{c}$ and $\tilde{\Delta}$ are skew-symmetric). Next, from (2.3) we have

$$
\begin{aligned}
\bar{\imath} & =x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\sum_{i j} \Delta_{i j} z_{i j}\left(x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right)_{i j} \\
& =x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\tilde{\Delta} \circ\left(x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right) \\
& =x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+x_{1}^{T} \tilde{\Delta} x_{2}-x_{2}^{T} \tilde{\Delta} x_{1} \\
& =x_{1}^{T} A x_{2}-x_{2}^{T} A x_{1} \\
& =\binom{x_{1}}{x_{2}}^{T}\left(\begin{array}{cc}
0 & A \\
-A & 0
\end{array}\right)\binom{x_{1}}{x_{2}},
\end{aligned}
$$

where the matrix

$$
\left(\begin{array}{cc}
0 & A  \tag{2.4}\\
-A & 0
\end{array}\right)
$$

is symmetric since $A$ is skew-symmetric. Denote by $\lambda$ the maximal eigenvalue of (2.4) (which is real), then from the above expression for $\bar{\imath}$ we have

$$
\begin{equation*}
\bar{\imath} \leq \lambda \tag{2.5}
\end{equation*}
$$

and there exists a vector $\binom{y_{1}}{y_{2}} \neq 0$ satisfying

$$
\left(\begin{array}{cc}
0 & A \\
-A & 0
\end{array}\right)\binom{y_{1}}{y_{2}}=\lambda\binom{y_{1}}{y_{2}}
$$

which implies

$$
A\left(y_{1}+y_{2} i\right)=-\lambda y_{2}+\lambda i y_{1}=\lambda i\left(y_{1}+y_{2} i\right),
$$

thus $\lambda i$ is an eigenvalue of $A$. Hence

$$
\lambda \leq \bar{\imath}
$$

by Theorem 1 , which combined with (2.5) gives

$$
\bar{\imath}=\lambda=\operatorname{Im}(\lambda i),
$$

hence $\bar{\imath}$ is achieved as the imaginary part of an eigenvalue of a matrix in $A^{I}$. To prove an analogous result for $\underline{i}$, let us apply the result just proved to the interval matrix $\left[-A_{c}-\Delta,-A_{c}+\Delta\right]$, which is also skew-symmetric. Then we have

$$
\operatorname{Im} \lambda=\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}^{T}-A_{c}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right)
$$

for an eigenvalue $\lambda$ of some $\tilde{A} \in\left[-A_{c}-\Delta, A_{c}+\Delta\right]$, hence

$$
\operatorname{Im}(-\lambda)=-\operatorname{Im} \lambda=\min _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}-\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right)=\underline{i}
$$

for the eigenvalue $-\lambda$ of $-\tilde{A} \in\left[A_{c}-\Delta, A_{c}+\Delta\right]$, which shows that $\underline{i}$ is achieved as well.

## 3 Practical bounds

Theorem 3 Let $A^{I}=\left[A_{c}-\Delta, A_{c}+\Delta\right]$ be a square interval matrix. Then for each eigenvalue $\lambda$ of each $A \in A^{I}$ we have

$$
\begin{gather*}
\lambda_{\min }\left(A_{c}^{\prime}\right)-\lambda_{\max }\left(\Delta^{\prime}\right) \leq \operatorname{Re} \lambda \leq \lambda_{\max }\left(A_{c}^{\prime}\right)+\lambda_{\max }\left(\Delta^{\prime}\right)  \tag{3.1}\\
\lambda_{\min }\left(A_{c}^{\prime \prime}\right)-\lambda_{\max }\left(\Delta^{\prime \prime}\right) \leq \operatorname{Im} \lambda \leq \lambda_{\max }\left(A_{c}^{\prime \prime}\right)+\lambda_{\max }\left(\Delta^{\prime \prime}\right), \tag{3.2}
\end{gather*}
$$

where

$$
\begin{aligned}
A_{c}^{\prime} & =\frac{1}{2}\left(A_{c}+A_{c}^{T}\right) \\
\Delta^{\prime} & =\frac{1}{2}\left(\Delta+\Delta^{T}\right) \\
A_{c}^{\prime \prime} & =\left(\begin{array}{cc}
0 & \frac{1}{2}\left(A_{c}-A_{c}^{T}\right) \\
\frac{1}{2}\left(A_{c}^{T}-A_{c}\right) & 0
\end{array}\right) \\
\Delta^{\prime \prime} & =\left(\begin{array}{cc}
0 & \Delta^{\prime} \\
\Delta^{\prime} & 0
\end{array}\right)
\end{aligned}
$$

Comments. $\lambda_{\min }, \lambda_{\max }$ denote the minimal and maximal eigenvalue of a symmetric matrix, respectively. Notice that all the matrices $A_{c}^{\prime}, \Delta^{\prime}, A_{c}^{\prime \prime}, \Delta^{\prime \prime}$ are symmetric by definition. Since $\lambda_{\max }(D)=\varrho(D)$ (spectral radius) holds for a nonnegative symmetric matrix $D$, the formulae (3.1), (3.2) may also be written in the form

$$
\begin{gathered}
\lambda_{\min }\left(A_{c}^{\prime}\right)-\varrho\left(\Delta^{\prime}\right) \leq \operatorname{Re} \lambda \leq \lambda_{\max }\left(A_{c}^{\prime}\right)+\varrho\left(\Delta^{\prime}\right) \\
\lambda_{\min }\left(A_{c}^{\prime \prime}\right)-\varrho\left(\Delta^{\prime \prime}\right) \leq \operatorname{Im} \lambda \leq \lambda_{\max }\left(A_{c}^{\prime \prime}\right)+\varrho\left(\Delta^{\prime \prime}\right)
\end{gathered}
$$

Proof. Let $\lambda$ be an eigenvalue of a matrix $A \in A^{I}$.

1) Since

$$
\begin{aligned}
\bar{r} & =\max _{\|x\|_{2}=1}\left(x^{T} A_{c} x+|x|^{T} \Delta|x|\right) \\
& \leq \max _{\|x\|_{2}=1} x^{T} A_{c} x+\max _{\|x\|_{2}=1}|x|^{T} \Delta|x| \\
& =\max _{\|x\|_{2}=1} x^{T} A_{c}^{\prime} x+\max _{\|x\|_{2}=1}|x|^{T} \Delta^{\prime}|x| \\
& =\lambda_{\max }\left(A_{c}^{\prime}\right)+\lambda_{\max }\left(\Delta^{\prime}\right),
\end{aligned}
$$

by Theorem 1 there holds

$$
\operatorname{Re} \lambda \leq \lambda_{\max }\left(A_{c}^{\prime}\right)+\lambda_{\max }\left(\Delta^{\prime}\right)
$$

which is the right-hand side inequality in (3.1).
2) The proof of the left-hand side inequality is analogous since

$$
\underline{r} \geq \min _{\|x\|_{2}=1} x^{T} A_{c} x-\max _{\|x\|_{2}=1}|x|^{T} \Delta|x|=\lambda_{\min }\left(A_{c}^{\prime}\right)-\lambda_{\max }\left(\Delta^{\prime}\right) .
$$

3) We have

$$
\begin{aligned}
\bar{\imath}= & \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}+\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right) \\
\leq & \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T} A_{c} x_{2}-x_{2}^{T} A_{c} x_{1}\right)+\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(\left|x_{1}\right|^{T} \Delta\left|x_{2}\right|+\left|x_{2}\right|^{T} \Delta\left|x_{1}\right|\right) \\
= & \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{x_{1}}{x_{2}}^{T}\left(\begin{array}{cc}
0 & A_{c} \\
-A_{c} & 0
\end{array}\right)\binom{x_{1}}{x_{2}}+\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{\left|x_{1}\right|}{\left|x_{2}\right|}^{T}\left(\begin{array}{ll}
0 & \Delta \\
\Delta & 0
\end{array}\right)\binom{\left|x_{1}\right|}{\left|x_{2}\right|} \\
= & \max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{x_{1}}{x_{2}}^{T}\left(\begin{array}{cc}
0 & \frac{1}{2}\left(A_{c}-A_{c}^{T}\right) \\
\frac{1}{2}\left(A_{c}^{T}-A_{c}\right) & 0
\end{array}\right)\binom{x_{1}}{x_{2}} \\
& +\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{\left|x_{1}\right|}{\left|x_{2}\right|}^{T}\left(\begin{array}{cc}
0 & \frac{1}{2}\left(\Delta+\Delta^{T}\right) \\
\frac{1}{2}\left(\Delta+\Delta^{T}\right) & 0
\end{array}\right)\binom{\left|x_{1}\right|}{\left|x_{2}\right|} \\
= & \lambda_{\max }\left(A_{c}^{\prime \prime}\right)+\lambda_{\max }\left(\Delta^{\prime \prime}\right) .
\end{aligned}
$$

Hence Theorem 1 gives

$$
\operatorname{Im} \lambda \leq \bar{\imath} \leq \lambda_{\max }\left(A_{c}^{\prime \prime}\right)+\lambda_{\max }\left(\Delta^{\prime \prime}\right)
$$

which is the right-hand side inequality in (3.2).
4) An analogous reasoning gives

$$
\begin{aligned}
\underline{i}= & \min _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\left(x_{1}^{T}\left(A_{c}-A_{c}^{T}\right) x_{2}-\Delta \circ\left|x_{1} x_{2}^{T}-x_{2} x_{1}^{T}\right|\right) \\
\geq & \min _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{x_{1}}{x_{2}}^{T}\left(\begin{array}{cc}
0 & \frac{1}{2}\left(A_{c}-A_{c}^{T}\right) \\
\frac{1}{2}\left(A_{c}^{T}-A_{c}\right) & 0
\end{array}\right)\binom{x_{1}}{x_{2}} \\
& -\max _{\left\|\left(x_{1}, x_{2}\right)\right\|_{2}=1}\binom{\left|x_{1}\right|}{\left|x_{2}\right|}^{T}\left(\begin{array}{cc}
0 & \frac{1}{2}\left(\Delta+\Delta^{T}\right) \\
\frac{1}{2}\left(\Delta+\Delta^{T}\right) & 0
\end{array}\right)\binom{\left|x_{1}\right|}{\left|x_{2}\right|} \\
= & \lambda_{\min }\left(A_{c}^{\prime \prime}\right)-\lambda_{\max }\left(\Delta^{\prime \prime}\right),
\end{aligned}
$$

which in view of Theorem 1 implies the left-hand side inequality in (3.2).

## 4 Consequences

We keep the notations $A_{c}^{\prime}, \Delta^{\prime}, A_{c}^{\prime \prime}, \Delta^{\prime \prime}$ introduced in Theorem 3.
Corollary 1 If

$$
\lambda_{\max }\left(A_{c}^{\prime}\right)+\lambda_{\max }\left(\Delta^{\prime}\right)<0
$$

then $A^{I}$ is (Hurwitz) stable.
Proof. Indeed, in this case Theorem 3 implies $\operatorname{Re} \lambda<0$ for each eigenvalue $\lambda$ of each $A \in A^{I}$.

We note that a symmetric interval matrix may contain nonsymmetric matrices with complex eigenvalues. Similarly, a skew-symmetric interval matrix may contain matrices with nonzero real parts.

Corollary 2 If $A^{I}$ is symmetric, then

$$
|\operatorname{Im} \lambda| \leq \lambda_{\max }\left(\Delta^{\prime \prime}\right)
$$

for each eigenvalue $\lambda$ of each $A \in A^{I}$.
Proof. The result follows from (3.2) since $A_{c}^{\prime \prime}=0$ in view of symmetry of $A_{c}$, hence $\lambda_{\min }\left(A_{c}^{\prime \prime}\right)=\lambda_{\max }\left(A_{c}^{\prime \prime}\right)=0$.

Corollary 3 If $A^{I}$ is skew-symmetric, then

$$
|\operatorname{Re} \lambda| \leq \lambda_{\max }\left(\Delta^{\prime}\right)
$$

for each eigenvalue $\lambda$ of each $A \in A^{I}$.
Proof. The assertion is a consequence of (3.1) since $A_{c}^{\prime}=0$.

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## Bibliography

[1] D. Hertz. The extreme eigenvalues and stability of real symmetric interval matrices. IEEE Transactions on Automatic Control, 37:532-535, 1992.
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