

# Nguyen Xuan Binh 887799 Exercise Sheet 5

Exercise 3: Let  $A, B \in \mathbb{R}^{n \times n}$  be symmetric. In addition, assume that

$$|\lambda_{\min}(A)| \geq 2 \text{ and } |\lambda_{\max}(B)| \leq 1$$

Show that

a)  $\|A^{-1}\|_2 \leq 1/2$

We have:  $\|A^{-1}\|_2 = \frac{1}{|\lambda_{\min}(A)|}$ . Since  $|\lambda_{\min}(A)| \geq 2$   
 $= \frac{1}{|\lambda_{\min}(A)|} \leq \frac{1}{2} \Rightarrow \|A^{-1}\|_2 \leq \frac{1}{2}$  (proven)

b) The matrix  $A + B$  is invertible

We have  $\|B\|_2 = |\lambda_{\max}(B)| \leq 1$   
 $\|A^{-1}\|_2 \leq \frac{1}{2} \Rightarrow \frac{1}{\|A^{-1}\|_2} \geq 2$

$$\Rightarrow \|B\|_2 \leq \frac{1}{\|A^{-1}\|_2}. \text{ Since eigenvalues of } A^{-1} \text{ and } B \neq \emptyset \Rightarrow A^{-1} \text{ and } B \text{ are invertible}$$

$\Rightarrow A + B$  is also invertible (proven)

c) Taking for granted the formula  $\|(A + B)^{-1}\|_2 \leq \frac{\|A^{-1}\|_2}{1 - \|B\|_2 \|A^{-1}\|_2}$

Prove a solution  $x$  to the equation  $(A + B)x = b$  satisfies  $\|x\|_2 \leq \|b\|_2$

We have:  $\frac{\|A^{-1}\|_2}{1 - \|B\|_2 \|A^{-1}\|_2} = \frac{1/|\lambda_{\min}(A)|}{1 - |\lambda_{\max}(B)| \frac{1}{|\lambda_{\min}(A)|}}$

$$= \frac{1}{|\lambda_{\min}(A)| - |\lambda_{\max}(B)|} \text{ As } |\lambda_{\min}(A)| \geq 2, |\lambda_{\max}(B)| \leq 1$$

$$\Rightarrow |\lambda_{\min}(A)| - |\lambda_{\max}(B)| \geq 1 \Rightarrow \frac{1}{|\lambda_{\min}(A)| - |\lambda_{\max}(B)|} \leq 1$$

$$\Rightarrow \|(A + B)^{-1}\|_2 \leq 1$$

We have:  $(A + B)x = b \Rightarrow x = (A + B)^{-1}b$

$$\Rightarrow \|x\|_2 = \|(A + B)^{-1}b\|_2 \leq \|(A + B)^{-1}\|_2 \|b\|_2$$

As  $\|(A + B)^{-1}\|_2 \leq 1 \Rightarrow \|(A + B)^{-1}\|_2 \|b\|_2 \leq \|b\|_2$  (Since 2 norm is strictly positive)

$$\Rightarrow \|x\|_2 = \|(A + B)^{-1}\|_2 \|b\|_2 \leq \|b\|_2 \Rightarrow \|x\|_2 \leq \|b\|_2 \text{ (proven)}$$

Exercise 4:

Let  $A = \begin{bmatrix} 1 + \epsilon^2 & 1 \\ 1 & 1 + \epsilon^2 \end{bmatrix}$  and  $b = \begin{bmatrix} 2 \\ 2 - \epsilon \end{bmatrix}$

a) Compute  $k_2(A) = \|A\|_2 \|A^{-1}\|_2$ . What happens to the condition number  $k_2(A)$  when  $\epsilon \rightarrow 0$ ?

We have:  $\det(A - \lambda I) = \begin{bmatrix} \epsilon^2 + 1 - \lambda & 1 \\ 1 & \epsilon^2 + 1 - \lambda \end{bmatrix}$

$\Rightarrow \det(A - \lambda I) = (\lambda - \epsilon^2)(\lambda - (\epsilon^2 + 2)) = 0$

$\Rightarrow \begin{cases} \lambda_1 = \epsilon^2 \\ \lambda_2 = \epsilon^2 + 2 \end{cases}$

$\square k_2(A) = \|A\|_2 \|A^{-1}\|_2 = \frac{|\lambda_{\max}(A)|}{|\lambda_{\min}(A)|} = \frac{\epsilon^2 + 2}{\epsilon^2} = 1 + \frac{2}{\epsilon^2}$

$\Rightarrow$  As  $\epsilon \rightarrow 0$ ,  $k_2(A) \rightarrow 1$

b) Compute  $R(A)$  and  $N(A)$  when  $\epsilon \neq 0$ . What about  $\epsilon = 0$ ?

For  $\epsilon \neq 0 \Rightarrow \lambda_1$  and  $\lambda_2$  are non-zero

As  $A$  has non-zero eigenvalues  $\Rightarrow A$  is an invertible matrix  $\Rightarrow A^{-1}$  exist

We have:  $R(A) = Ay$  for all  $y \in \mathbb{R}^n$ . As  $A^{-1}$  exist, let  $x$  be a specific solution

$\Rightarrow x = Ay \Rightarrow A^{-1}x = A^{-1}Ay = A^{-1}x = y$

$\Rightarrow$  For all  $x \in \mathbb{R}^2$  there always exist a vector  $y$  so that  $A^{-1}x = y$

$\Rightarrow R(A) = \mathbb{R}^{2 \times 2}$  for all possible matrices in dimension  $\mathbb{R}^{2 \times 2}$

\* Null space:  $Ax = 0 \Rightarrow A^{-1}Ax = A^{-1} \cdot 0 \Rightarrow x = 0$

$\Rightarrow x = 0$  is the unique solution

$\Rightarrow N(A)$  has only trivial solution  $x = 0$

$\square$  For  $\epsilon = 0 \Rightarrow \lambda_1$  is a zero eigenvalue  $\Rightarrow A$  is a non-invertible matrix

$\Rightarrow A^{-1}$  doesn't exist

$\epsilon = 0 \Rightarrow A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ . Let  $y = [y_1 \ y_2]^T$  be an arbitrary vector

$\Rightarrow Ay = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} y_1 + y_2 \\ y_1 + y_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = y^T \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$\Rightarrow R(A) = \left\{ y^T \begin{bmatrix} 1 \\ 1 \end{bmatrix}, y \in \mathbb{R}^{1 \times 2} \right\}$

\* Null space:  $Ax = 0 \Rightarrow \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \Rightarrow \begin{cases} x_1 + x_2 = 0 \\ x_1 + x_2 = 0 \end{cases} \Rightarrow x_1 = -x_2$

$\Rightarrow N(A) = \left\{ [1 - 1]^s \cup [-1 1]^t \mid s, t \in \mathbb{R} \right\}$

c) For  $\epsilon \neq 0$ , solve  $Ax = b$  using the Cramer's rule

$$A^{-1} = \frac{1}{\det A} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}$$

We have:  $Ax = b \Rightarrow \begin{bmatrix} 1 + \epsilon^2 & 1 \\ 1 & 1 + \epsilon^2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 - \delta \end{bmatrix}$

(Since it is proved before that  $\epsilon \neq 0$  means  $N(A)$  is trivial, there's no need to find  $N(A)$ )

We have:  $A^{-1} = \frac{1}{(\epsilon^2 + 1)^2 - 1} \begin{bmatrix} 1 + \epsilon^2 & -1 \\ -1 & 1 + \epsilon^2 \end{bmatrix}$

$$\Rightarrow A^{-1}Ax = A^{-1}b \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{1}{(\epsilon^2 + 1)^2 - 1} \begin{bmatrix} \epsilon^2 + 1 & -1 \\ -1 & \epsilon^2 + 1 \end{bmatrix} \begin{bmatrix} 2 \\ 2 - \delta \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{1}{(\epsilon^2 + 1)^2 - 1} \begin{bmatrix} 2\epsilon^2 + \delta \\ 2\epsilon^2 - \delta\epsilon^2 - \delta \end{bmatrix} \text{ (answer)}$$

When  $\delta = 0 \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \frac{1}{(\epsilon^2 + 1)^2 - 1} \begin{bmatrix} 2\epsilon^2 \\ 2\epsilon^2 \end{bmatrix} = \left[ \frac{2\epsilon^2}{\epsilon^4 + 2\epsilon^2}, \frac{2\epsilon^2}{\epsilon^4 + 2\epsilon^2} \right]^T$

$$\Rightarrow \bar{x} = \left[ \frac{2}{\epsilon^2 + 2}, \frac{2}{\epsilon^2 + 2} \right]^T \Rightarrow \lim_{\epsilon \rightarrow 0} \bar{x} = \left[ \frac{2}{0^2 + 2}, \frac{2}{0^2 + 2} \right]^T = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

When  $\delta = 0$ , when  $\epsilon \rightarrow 0 \Rightarrow \bar{x} \rightarrow [1 \ 1]^T$

When  $\delta \neq 0 \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \left[ \frac{2\epsilon^2 + \delta}{\epsilon^4 + 2\epsilon^2}, \frac{2\epsilon^2 - \delta\epsilon^2 - \delta}{\epsilon^4 + 2\epsilon^2} \right]^T$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \left[ \frac{2}{\epsilon^2 + 2} + \frac{\delta}{\epsilon^4 + 2\epsilon^2}, \frac{2}{\epsilon^2 + 2} - \frac{\delta}{\epsilon^2 + 2} - \frac{\delta}{\epsilon^4 + 2\epsilon^2} \right]$$

$\epsilon \rightarrow 0^+$	$\downarrow$	$\downarrow$	$\downarrow$	$\downarrow$
1	1	$+\infty$	1	$-\delta/2$
$\epsilon \rightarrow 0^-$	1	$-\infty$	1	$-\delta/2$
				$+\infty$

$\Rightarrow$  When  $\delta \neq 0$ ,  $\epsilon \rightarrow 0^+ \Rightarrow \bar{x} \rightarrow [+\infty \ -\infty]^T$

$\epsilon \rightarrow 0^- \Rightarrow \bar{x} \rightarrow [-\infty \ +\infty]^T$