

Numerical Linear Algebra

Tamar Kldiashvili

Kutaisi International University

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► Quiz Solutions

Example 7.1

Prove if statement below is true, else give counter example:

1. $\text{cond}(A^{-1}) = \text{cond}(\alpha A)$, $A \in \mathbb{R}^{n \times n}$, $\alpha \in \mathbb{R}$, $\alpha \neq 0$;
2. $\text{cond}(A) = 1$ if and only if $A^T A = \alpha I$, $\alpha \in \mathbb{R}$, $\alpha \neq 0$;
3. $\text{cond}_2(A) = \text{cond}_2(A^T)$ iff $A \in \mathbb{R}^{n \times n}$;
4. $\text{cond}_\infty(A) = \text{cond}_1(A)$ iff A is symmetric.

Solution 7.1

1. $\text{cond}(A^{-1}) = \text{cond}(\alpha A)$, $A \in \mathbb{R}^{n \times n}$, $\alpha \in \mathbb{R}$, $\alpha \neq 0$;

Solution:

$$\begin{aligned}\text{cond}(\alpha A) &= \|\alpha A\| \cdot \|(\alpha A)^{-1}\| = |\alpha| \cdot \|A\| \cdot |\alpha^{-1}| \cdot \|A^{-1}\| = \\ &= \|A\| \cdot \|A^{-1}\| = \text{cond}(A^{-1})\end{aligned}$$

Solution 7.1

2. $\text{cond}(A) = 1$ if and only if $A^T A = \alpha I$, $\alpha \in \mathbb{R}, \alpha \neq 0$;

Solution:

Consider $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$.

$$A^T A = A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = 1 \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

But $\text{cond}_F(A) = \sqrt{2}$.

Solution 7.1

3. $\text{cond}_2(A) = \text{cond}_2(A^T)$ iff $A \in \mathbb{R}^{n \times n}$;

Solution:

I. Prove $\text{cond}_2(A) = \text{cond}_2(A^T) \Rightarrow A \in \mathbb{R}^{n \times n}$

$$\text{cond}_2(A) = \|A\|_2 \|A^{-1}\|_2$$

The inverse matrix exist only for square matrices. So, $A \in \mathbb{R}^{n \times n}$.

II. Prove $A \in \mathbb{R}^{n \times n} \Rightarrow \text{cond}_2(A) = \text{cond}_2(A^T)$.

Matrix and its transpose have the same set of eigenvalues. Thus,

$$\text{cond}_2(A) = \frac{\sigma_1}{\sigma_2} = \text{cond}_2(A^T)$$

Solution 7.1

4. $\text{cond}_\infty(A) = \text{cond}_1(A)$ iff A is symmetric.

Solution:

Consider $A = \begin{pmatrix} 1 & 2 & 0 \\ 1 & 1 & 1 \\ 1 & 0 & 0 \end{pmatrix}$.

$$A^{-1} = \begin{pmatrix} 0 & 0 & 1 \\ 1/2 & 0 & -1/2 \\ -1/2 & 1 & -1/2 \end{pmatrix}.$$

$$\text{cond}_\infty(A) = \|A\|_\infty \cdot \|A^{-1}\|_\infty = 3 \cdot 2 = 6$$

$$\text{cond}_1(A) = \|A\|_1 \cdot \|A^{-1}\|_1 = 3 \cdot 2 = 6$$

$\text{cond}_\infty(A) = \text{cond}_1(A)$ but A is not symmetric.

Example 7.2

Suppose

$$Ax = b, A = \begin{pmatrix} 0.99 & 1 & 1 \\ 1 & 0.9999 & 1 \\ 1 & 1 & 0.999999 \end{pmatrix}, \text{cond}_1(A) \approx 59991, \|b\|_1 = \|A\|_1$$

What is acceptable absolute error in the right hand side and coefficients of linear system $Ax = b$ for obtaining solution with relative error not higher than 10^{-3} ?

Example 7.3

Give example of 2×2 matrix such that $\text{cond}_2(A) = 10^9$.

Solution:

Use the fact that $\text{cond}_2(A) = \frac{\lambda_{\max}}{\lambda_{\min}}$, where A is a symmetric positive definite matrix.

We want $\text{cond}_2(A) = 10^9 \Rightarrow \text{cond}_2(A) = \frac{\lambda_{\max}}{\lambda_{\min}} = 10^9$.

If we take $A = \begin{pmatrix} 10^9 & 0 \\ 0 & 1 \end{pmatrix}$, we get that $\lambda_{\max} = 10^9$ and $\lambda_{\min} = 1$. Thus,
 $\text{cond}_2(A) = \frac{\lambda_{\max}}{\lambda_{\min}} = 10^9$

Example 7.4

Suppose $v_1 = (1, 1, 1, 1)$, $v_2 = (2, 2, 2, 2)$, $v_3 = (1.5, 0.5, 1.5, 0.5)$. Using different markers for different vectors draw points (i, v_{1i}) , (i, v_{2i}) , (i, v_{3i}) , $i = 1, 2, 3, 4$ on a plane. Give example of a vector norm formula such that:

1. $\|v_2 - v_1\| < \|v_3 - v_1\|$
2. $\|v_3 - v_1\| < \|v_2 - v_1\|$

Solution 7.4

$$1. \quad \|v_2 - v_1\| < \|v_3 - v_1\|$$

Solution:

$$(v_2 - v_1) = (1, 1, 1, 1)$$

$$(v_3 - v_1) = (0.5, -0.5, 0.5, -0.5)$$

Check if $\|x\|_* = |x_1| + \sum_{i=2}^n |x_i - x_{i-1}|$ is a norm.

1. $\|x\|_* > 0$ when $x \neq 0$ and $\|x\|_* = 0$ iff $x = 0$;
2. $\|kx\|_* = |kx_1| + \sum_{i=2}^n |kx_i - kx_{i-1}| = |k||x_1| + |k| \sum_{i=2}^n |x_i - x_{i-1}| = |k|(|x_1| + \sum_{i=2}^n |x_i - x_{i-1}|) = |k|\|x\|_*$ for any scalar k .
3. $\|x + y\|_* = |x_1 + y_1| + \sum_{i=2}^n |(x_i + y_i) - (x_{i-1} + y_{i-1})| \leq |x_1| + |y_1| + \sum_{i=2}^n (|x_i - x_{i-1}| + |y_i - y_{i-1}|)$ Thus, $\|x\|_*$ is a norm.

Check if given inequality is satisfied.

$$\|v_2 - v_1\|_* = 1 + 0 + 0 + 0 = 1 < 0.5 + 1 + 1 + 1 = 3.5 = \|v_3 - v_1\|_*$$

Solution 7.4

$$2. \quad \|v_3 - v_1\| < \|v_2 - v_1\|$$

Solution:

$$(v_2 - v_1) = (1, 1, 1, 1)$$

$$(v_3 - v_1) = (0.5, -0.5, 0.5, -0.5)$$

Consider 2-norm: $\|x\|_2 = (\sum_{i=1}^n |x_i|^2)^{1/2}$

$$\|v_3 - v_1\|_2 = \sqrt{1/4 + 1/4 + 1/4 + 1/4} = 1 < \sqrt{1 + 1 + 1 + 1} = 2 = \|v_2 - v_1\|_2$$

Example 7.5

Consider $n \times n$ bi-diagonal matrix with ones on main diagonal and twos on upper diagonal. Show that this matrix is ill-conditioned for $n = 101$.

Example 7.6

Prove: $\|A\| = \inf\{\lambda \in \mathbb{R} : \|Ax\| \leq \lambda\|x\|, x \in \mathbb{R}^n\}$, $A \in \mathbb{R}^{n \times n}$