

2.9 Prove that, for any nonsingular matrix $A \in \mathbb{R}^{n \times n}$,

$$\kappa_2(A) = \left(\frac{\lambda_n}{\lambda_1} \right)^{1/2},$$

where λ_1 is the smallest and λ_n is the largest eigenvalue of the matrix $A^T A$.

Show that the condition number $\kappa_2(Q)$ of an orthogonal matrix Q is equal to 1. Conversely, if $\kappa_2(A) = 1$ for the matrix A , show that all the eigenvalues of $A^T A$ are equal; deduce that A is a scalar multiple of an orthogonal matrix.

By definition $\kappa_2(A) = \|A\|_2 \|A^{-1}\|_2$. Now by

Theorem 2.9 Let $A \in \mathbb{R}^{n \times n}$ and denote the eigenvalues of the matrix $B = A^T A$ by λ_i , $i = 1, 2, \dots, n$. Then,

$$\|A\|_2 = \max_{i=1}^n \lambda_i^{1/2}.$$

we have that

$$\begin{aligned} \|A\|_2 &= \max \{ \|Ax\|_2 : \|x\|_2 = 1 \} \\ &= \max \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } A^T A \} = \lambda_n^{1/2} \end{aligned}$$

and similarly that

$$\begin{aligned} \|A^{-1}\|_2 &= \max \{ \|A^{-1}x\|_2 : \|x\|_2 = 1 \} \\ &= \max \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } (A^{-1})^T A^{-1} \}. \end{aligned}$$

Since $(A^{-1})^T = (A^T)^{-1}$ then $(A^{-1})^T A^{-1} = (A^T)^{-1} A^{-1} = (AA^T)^{-1} = C^{-1}$ where $C = AA^T$. Suppose λ and x are an eigenvalue-eigenvector

pair such that $Cx = \lambda x$. Then multiplying by C^{-1} yields that $x = C^{-1}\lambda x = \lambda C^{-1}x$ or equivalently

$$C^{-1}x = \frac{1}{\lambda}x.$$

Therefore the eigenvalues of C^{-1} are given by $\frac{1}{\lambda}$ where λ is an eigenvalue of C . It follows that

$$\begin{aligned}\|A^{-1}\|_2 &= \max \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } (AA^T)^{-1} \} \\ &= \frac{1}{\min \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } AA^T \}}.\end{aligned}$$

To finish we need show that $C = AA^T$ and $B = A^T A$ have the same eigenvalues.

Suppose $Bx = \lambda x$. Define $y = Ax$. Then

$$Cy = CAx = AA^T Ax = ABx = A\lambda x = \lambda Ax = \lambda y$$

shows that λ is an eigenvalue of C .

Similarly if $Cy = \lambda y$ then defining $x = A^T y$ yields

$$Bx = BA^T y = A^T AA^T y = A^T Cy = A^T \lambda y = \lambda A^T y = \lambda x$$

and so λ is an eigenvalue of B . It follows that

$$\|A^{-1}\|_2 = \frac{1}{\min \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } A^T A \}} = \frac{1}{\lambda_1^{1/2}}$$

Consequently $\kappa_2(A) = \frac{\lambda_n^{1/2}}{\lambda_1^{1/2}} = \left(\frac{\lambda_n}{\lambda_1}\right)^{1/2}$ as desired.

2.10 Let $A \in \mathbb{R}^{n \times n}$. Show that if λ is an eigenvalue of $A^T A$, then

$$0 \leq \lambda \leq \|A^T\| \|A\|,$$

provided that the same subordinate matrix norm is used for both A and A^T . Hence show that, for any nonsingular $n \times n$ matrix A ,

$$\kappa_2(A) \leq \{\kappa_1(A) \kappa_\infty(A)\}^{1/2}.$$

Let λ and ξ be an eigenvalue-eigenvector pair such that

$$A^T A \xi = \lambda \xi.$$

Since $A^T A \xi \cdot \xi = A \xi \cdot A \xi = \|A \xi\|_2^2$ and $\lambda \xi \cdot \xi = \lambda \|\xi\|_2^2$ it follows that

$$\lambda = \frac{\|A \xi\|_2^2}{\|\xi\|_2^2} \geq 0$$

Now for any vector norm $\|\cdot\|$ and subordinate matrix norm $\|\cdot\|$ holds

$$\lambda \|\xi\| = |\lambda| \|\xi\| = \|\lambda \xi\| = \|A^T A \xi\| \leq \|A^T\| \|A\| \|\xi\|.$$

Thus $\lambda \leq \|A^T\| \|A\|$ and so $0 \leq \lambda \leq \|A^T\| \|A\|$.

We know that $\|A\|_1 = \max_{j=1}^n \sum_{i=1}^n |a_{ij}|$ and $\|A\|_\infty = \max_{i=1}^n \sum_{j=1}^n |a_{ij}|$.

It follows that

$$\|A^T\|_1 = \max_{j=1}^n \sum_{i=1}^n |[A^T]_{ij}| = \max_{j=1}^n \sum_{i=1}^n |a_{ji}| = \max_{i=1}^n \sum_{j=1}^n |a_{ij}| = \|A\|_\infty$$

and

$$\|A^T\|_\infty = \max_{i=1}^n \sum_{j=1}^n |[A^T]_{ij}| = \max_{i=1}^n \sum_{j=1}^n |a_{ji}| = \max_{j=1}^n \sum_{i=1}^n |a_{ij}| = \|A\|_1$$

In particular $\|A^T\|_1 = \|A\|_\infty$ and $\|A^T\|_\infty = \|A\|_1$.

Now

$$\|A\|_2 = \max \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } A^T A \} \leq \|A^T\|_\infty^{1/2} \|A\|_\infty^{1/2} = \|A\|_1^{1/2} \|A\|_\infty^{1/2}$$

Similarly

$$\begin{aligned} \|A^{-1}\|_2 &= \max \{ \lambda^{1/2} : \lambda \text{ is an eigenvalue of } (A^{-1})^T (A^{-1}) \} \leq \|(A^{-1})^T\|_\infty^{1/2} \|A^{-1}\|_\infty^{1/2} \\ &= \|A^{-1}\|_1^{1/2} \|A^{-1}\|_\infty^{1/2} \end{aligned}$$

It follows that

$$\begin{aligned} \kappa_2(A) &= \|A\|_2 \|A^{-1}\|_2 \leq \|A\|_1^{1/2} \|A\|_\infty^{1/2} \|A^{-1}\|_1^{1/2} \|A^{-1}\|_\infty^{1/2} \\ &= \|A\|_1^{1/2} \|A^{-1}\|_1^{1/2} \|A\|_\infty^{1/2} \|A^{-1}\|_\infty^{1/2} = \kappa_1(A)^{1/2} \kappa_\infty(A)^{1/2} \end{aligned}$$

2.12 Let $B \in \mathbb{R}^{n \times n}$ and denote by I the identity matrix of order n . Show that if the matrix $I - B$ is singular, then there exists a nonzero vector $x \in \mathbb{R}^n$ such that $(I - B)x = \mathbf{0}$; deduce that $\|B\| \geq 1$, and hence that, if $\|A\| < 1$, then the matrix $I - A$ is nonsingular.

Now suppose that $A \in \mathbb{R}^{n \times n}$ with $\|A\| < 1$. Show that

$$(I - A)^{-1} = I + A(I - A)^{-1},$$

and hence that

$$\|(I - A)^{-1}\| \leq 1 + \|A\| \|(I - A)^{-1}\|.$$

Deduce that

$$\|(I - A)^{-1}\| \leq \frac{1}{1 - \|A\|}.$$

If $I - B$ is singular then the nullspace of $I - B$ is non-trivial and

so there is $x \neq 0$ such that $(I-B)x=0$. Then $x=Bx$ and so

$$\|x\| = \|Bx\| \leq \|B\| \|x\|$$

implies after cancelling by $\|x\|$ that $\|B\| \geq 1$. Consequently, if $\|A\| < 1$ then $I-A$ is non-singular.

Now, multiplying

$$(I-A)(I+A(I-A)^{-1}) = I-A + (I-A)A(I-A)^{-1}$$

$$= I-A + (A-A^2)(I-A)^{-1} = I-A + A(I-A)(I-A)^{-1}$$

$$= I-A + AI = I-A+A = I$$

Shows that $I+A(I-A)^{-1}$ is an inverse of $(I-A)$. Since inverses are unique, it follows that $(I-A)^{-1} = I+A(I-A)^{-1}$. Thus

$$\|(I-A)^{-1}\| \leq \|I\| + \|A(I-A)^{-1}\| \leq 1 + \|A\| \|(I-A)^{-1}\|$$

implies

$$(1 - \|A\|) \|(I-A)^{-1}\| \leq 1.$$

Therefore

$$\|(I-A)^{-1}\| \leq \frac{1}{1 - \|A\|}.$$