

2. SOLUTIONS

(1) Problem 1:

Proof. (a) If $A = 0$, then clearly $Ax = 0, \forall x \in R^n$, so

$$\|A\| = \max_{x \in R^n, x \neq 0} \frac{\|Ax\|}{\|x\|} = 0.$$

Otherwise $\exists a_{ij} \neq 0, a_{ij} \in A$. Let $e_j = [0, \dots, 1, \dots, 0]^T$, then

$$\|A\| = \max_{x \in R^n, x \neq 0} \frac{\|Ax\|}{\|x\|} \geq \frac{\|Ae_j\|}{\|e_j\|} > 0.$$

In summary, $\|A\| \geq 0$, and $\|A\| = 0$ iff $A = 0$.

(b)

$$\begin{aligned} \|\alpha A\| &= \max_{x \in R^n, x \neq 0} \frac{\|\alpha Ax\|}{\|x\|} = \max_{x \in R^n, x \neq 0} \frac{|\alpha| \|Ax\|}{\|x\|} \\ &= |\alpha| \max_{x \in R^n, x \neq 0} \frac{\|Ax\|}{\|x\|} = |\alpha| \|A\| \end{aligned}$$

(c)

$$\begin{aligned} \|A + B\| &= \max_{x \in R^n, x \neq 0} \frac{\|(A + B)x\|}{\|x\|} = \max_{x \in R^n, x \neq 0} \frac{\|Ax + Bx\|}{\|x\|} \\ &\leq \max_{x \in R^n, x \neq 0} \frac{\|Ax\| + \|Bx\|}{\|x\|} \\ &\leq \max_{x \in R^n, x \neq 0} \frac{\|Ax\|}{\|x\|} + \max_{x \in R^n, x \neq 0} \frac{\|Bx\|}{\|x\|} = \|A\| + \|B\| \end{aligned}$$

(d) According to the definition of the matrix norm, $\|A\| = \max_{x \in R^n} \|Ax\|/\|x\|$ we have $\|Ax\| \leq \|A\|\|x\|, \forall x \in R^n$. So, $\forall x \in R^n$ we have the following inequalities:

$$\|ABx\| \leq \|A\|\|Bx\| \leq \|A\|\|B\|\|x\|$$

$$\text{Therefore, } \|AB\| = \max_{x \in R^n, x \neq 0} \frac{\|ABx\|}{\|x\|} \leq \max_{x \in R^n, x \neq 0} \frac{\|A\|\|B\|\|x\|}{\|x\|} = \|A\|\|B\|.$$

□

(2) Problem 2:

Proof. $A \in R^{n \times n}$ is symmetric, so \exists a real diagonal matrix Λ and an orthonormal matrix Q such that $A = Q\Lambda Q^T$, where the diagonal values of Λ are A 's eigenvalues $\lambda_i, i = 1, \dots, n$ and the columns of Q are the corresponding eigenvectors.

According to Gerschgorin's theorem, A 's eigenvalues are located in the union of disks

$$d_i = \{z \in C : |z - a_{i,i}| \leq \sum_{j \neq i} |a_{i,j}|\}, \quad i = 1, \dots, n.$$

Because A is positive and strictly row-wise diagonal dominant, so,

$$d_i = \{z \in \mathbb{C} : 0 < a_{i,i} - \sum_{j \neq i} a_{i,j} \leq z \leq \sum_j a_{i,j}\}, \quad i = 1, \dots, n.$$

Therefore all A 's eigenvalues are positive, then $\forall x \in \mathbb{R}^n, x \neq 0$ we have,

$$x^T A x = x^T Q \Lambda Q^T x = y^T \Lambda y > 0, \text{ where } y = Q^T x \neq 0.$$

i.e. A is positive definite. \square

- (3) Problem 3: First, recall that the set of complex numbers $\sigma(A) = \{\lambda : \det(A - \lambda I) = 0\}$ is called spectrum of A .

Proof. Then, $\forall \lambda \in \sigma(A)$, we show that $\lambda^k \in \sigma(A^k)$. Indeed, let $Au = \lambda u$ with $u \neq 0$.

$$A^k u = \lambda A^{k-1} u = \dots = \lambda^k u.$$

For any $\mu \in \sigma(A^k)$, we have $A^k v = \mu v, v \neq 0$. Since, v belong to the range of A , so at least there exists one eigenvalue $\lambda = \mu^{1/k}$ s.t.

$$Av = \mu^{1/k} v.$$

Thus, we always have that for any $|\mu|$ where $A^k u = \mu u, u \neq 0$, there exists a $|\lambda| = |\mu|^{1/k}$ where $Au = \lambda u, u \neq 0$; for any $|\lambda|$ where $Au = \lambda u, u \neq 0$, there exists a $|\mu| = |\lambda^k| = |\lambda|^k$ where $A^k u = \mu u, u \neq 0$. So,

$$\rho(A^k) = \max_{\mu \in \sigma(A^k)} |\mu| \equiv \max_{\lambda \in \sigma(A)} |\lambda|^k = \left(\max_{\lambda \in \sigma(A)} |\lambda| \right)^k = (\rho(A))^k.$$

\square

- (4) Problem 4:

Proof. It follows from the obvious string of inequalities

$$\begin{aligned} \|A\|_2 &= \max_{\|x\|_2=1} \|Ax\|_2 = \max_{\|x\|_2=1} \left(\sum_{i=1}^n \left(\sum_{j=1}^n a_{ij} x_j \right)^2 \right)^{1/2} \\ &\leq \max_{\|x\|_2=1} \left(\sum_{i=1}^n \sum_{j=1}^n a_{ij}^2 \sum_{j=1}^n x_j^2 \right)^{1/2} \quad \text{Schwartz inequality} \\ &= \left(\sum_{i=1}^n \sum_{j=1}^n a_{ij}^2 \right)^{1/2} \leq \left(\sum_{i=1}^n \left(\sum_{j=1}^n |a_{ij}| \right)^2 \right)^{1/2} \\ &\leq \left(n \left(\max_i \sum_{j=1}^n |a_{ij}| \right)^2 \right)^{1/2} \leq \left(n \|A\|_\infty^2 \right)^{1/2} = \sqrt{n} \|A\|_\infty \end{aligned}$$

\square

- (5) Problem 5:

Proof. Assume the norm $\|\cdot\|$ satisfies the submultiplicative property, i.e. $\|AB\| \leq \|A\| \|B\|$ (subordinate matrix norms have this property).

$$\begin{aligned} \|B^{-1} - A^{-1}\| &= \|A^{-1} - B^{-1}\| = \|B^{-1}(B - A)A^{-1}\| \\ &\leq \|B^{-1}\| \|B - A\| \|A^{-1}\| \end{aligned}$$

A, B are nonsingular matrices, divide $\|B^{-1}\|$ on both sides,

$$\begin{aligned} \frac{\|B^{-1} - A^{-1}\|}{\|B^{-1}\|} &\leq \|B - A\| \|A^{-1}\| = \frac{\|B - A\| \|A^{-1}\| \|A\|}{\|A\|} \\ &= \text{cond}(A) \frac{\|B - A\|}{\|A\|} \end{aligned}$$

□

(6) Problem 6:

Proof. For any symmetric matrix A , all its eigenvalues are real.

According to Gerschgorin's theorem, A 's eigenvalues are located in the union of disks

$$d_i = \{z \in C : |z - a_{i,i}| \leq \sum_{j \neq i} |a_{i,j}|\}, \quad i = 1, \dots, n.$$

i.e.,

$$d_i = \{z \in C : a_{i,i} - \sum_{j \neq i} |a_{i,j}| \leq z \leq a_{i,i} + \sum_{j \neq i} |a_{i,j}|\}, \quad i = 1, \dots, n.$$

Because A 's diagonal elements are positive and A is strictly diagonal dominant, so,

$$a_{i,i} - \sum_{j \neq i} |a_{i,j}| = |a_{i,i}| - \sum_{j \neq i} |a_{i,j}| > 0$$

$$a_{i,i} + \sum_{j \neq i} |a_{i,j}| = |a_{i,i}| + \sum_{j \neq i} |a_{i,j}| > 0, \quad i = 1, \dots, n.$$

So, all these disks are located at the right side of the y -axis in the complex plane, so all eigenvalues of A are positive. □

(7) Problem 7:

Proof. First possible solution:

- (1) Consider an eigenvalue λ and its eigenvector ψ of the matrix AB , that is $AB\psi = \lambda\psi$. We do not know whether λ and ψ are real, so we assume that they are complex. Recall that the inner product of two complex vectors ψ and ϕ is defined as $(\phi, \psi) = \sum_i \phi_i \bar{\psi}_i$, where $\bar{\psi}$ is the complex conjugate to ψ . Recall, that $(\phi, \psi) = \overline{(\psi, \phi)}$. For the inner product of the complex vectors $AB\psi$ and ψ we get

$$AB\psi = \lambda\psi \quad \Rightarrow \quad BAB\psi = \lambda B\psi, \quad \Rightarrow \quad (BAB\psi, \psi) = \lambda(B\psi, \psi).$$

Now using the symmetry of A and B we get

$$(BAB\psi, \psi) = (\psi, BAB\psi) = \overline{(BAB\psi, \psi)},$$

which means that the complex number $(BAB\psi, \psi)$ is equal to its complex conjugate, i.e. the number is real. In the same way we prove that $(B\psi, \psi)$ is real as well, from where we conclude that λ is real. Then we conclude that ψ is real as well.

- (2) If A and B are positive definite, then $(B\psi, \psi) > 0$ and $(BAB\psi, \psi) > 0$, therefore from $(BAB\psi, \psi) = \lambda(B\psi, \psi)$ it follows that $\lambda > 0$.

Another possible solution (for those with more advanced knowledge in linear algebra):

(1) A is SPD, its square root exists $A^{1/2}$ which is SPD, then

$$AB \sim A^{-1/2}ABA^{1/2} = A^{1/2}BA^{1/2}$$

Because both $A^{1/2}$ and B are symmetric, then

$$(A^{1/2}BA^{1/2})^T = (A^{1/2})^T B^T (A^{1/2})^T = A^{1/2}BA^{1/2},$$

so $A^{1/2}BA^{1/2}$ is symmetric, and the eigenvalues of $A^{1/2}BA^{1/2}$ are real. $AB \sim A^{1/2}BA^{1/2}$, they have the same spectrum, so all eigenvalues of AB are real.

(2) If B is also SPD, we can show that $A^{1/2}BA^{1/2}$ is SPD. The symmetry is shown in (1). Now show positive definite. $\forall x \in \mathbb{R}^n, x \neq 0$ with $y = A^{1/2}x \neq 0$ we have

$$(A^{1/2}BA^{1/2}x, x) = (BA^{1/2}x, A^{1/2}x) = (By, y) > 0.$$

So $A^{1/2}BA^{1/2}$ is SPD, and all its eigenvalues are positive. $AB \sim A^{1/2}BA^{1/2}$, they have the same spectrum, so all eigenvalues of AB are positive. \square

we show (b) as well

(8) Problem 8:

Proof. We prove (a). First we show that $\|A\|_\infty \leq \max_i \sum_j |a_{i,j}|$. Indeed,

$$\begin{aligned} \|A\|_\infty &= \max_{\|x\|_\infty=1} \|Ax\|_\infty \\ &= \max_{\|x\|_\infty=1} \max_i \left| \sum_{j=1}^n a_{ij}x_j \right| \\ &\leq \max_{\|x\|_\infty=1} \max_i \sum_{j=1}^n |a_{ij}| |x_j| \\ &\leq \max_{\|x\|_\infty=1} \max_i \max_j |x_j| \sum_{j=1}^n |a_{ij}| \\ &= \max_i \sum_{j=1}^n |a_{ij}| \end{aligned}$$

Next we show also that $\max_i \sum_j |a_{i,j}| \leq \|A\|_\infty$. Assume that for some integer $k \in [1, n]$ we have

$$\sum_{j=1}^n |a_{kj}| = \max_i \sum_{j=1}^n |a_{ij}|.$$

By choosing a vector y s.t.,

$$y_j = \begin{cases} 1, & \text{for } a_{k,j} \geq 0; \\ -1, & \text{for } a_{k,j} < 0; \end{cases}$$

Then, $\|y\|_\infty = 1$ and

$$\|A\|_\infty = \sup_{x \in \mathcal{R}^n, \|x\|_\infty=1} \|Ax\|_\infty \geq \|Ay\|_\infty = \sum_j |a_{k,j}| = \max_i \sum_j |a_{i,j}|.$$

Thus, from the inequalities $\max_i \sum_j |a_{i,j}| \leq \|A\|_\infty \leq \max_i \sum_j |a_{i,j}|$, it follows that

$$\|A\|_\infty = \max_i \sum_j |a_{i,j}|.$$

The inequality (b) is shown in a similar way. □