

1. For any $u \in \mathbf{R}^n$, let $|u|$ denote the (usual notion of) length of u ; and for any $A \in \mathbf{R}^{n \times n}$, let

$\|A\|$ denote the maximum value of $f(x_1, \dots, x_n) = \left| A \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \right|$ subject to the constraint

$g(x_1, \dots, x_n) = \left| \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \right| = 1$. Using only logic only (no calculations!), show that each

of the following properties hold:

- (a) $\|A + B\| \leq \|A\| + \|B\|$, for all $A, B \in \mathbf{R}^{n \times n}$.
 - (b) $\|kA\| = |k| \|A\|$, for all $A \in \mathbf{R}^{n \times n}$ and all $k \in \mathbf{R}$.
 - (c) $\|A\| = 0$ implies A is the zero matrix.
 - (d) $\|AB\| \leq \|A\| \|B\|$, for all $A, B \in \mathbf{R}^{n \times n}$.
 - (e) $\|I\| = 1$.
2. Let $A \in \mathbf{R}^{2 \times 2}$. We will use the method of Lagrange multipliers to maximise $\phi(x, y) = \left| A \begin{pmatrix} x \\ y \end{pmatrix} \right|^2$ subject to the constraint $\gamma(x, y) = \left| \begin{pmatrix} x \\ y \end{pmatrix} \right|^2 = 1$. [The squaring is just to get rid of the square roots, which makes calculating derivatives easier!]

- (a) Show that $\nabla \phi(x, y) = 2A^T A \begin{pmatrix} x \\ y \end{pmatrix}$.
- (b) Show that λ is a Lagrange multiplier if and only if it is an eigenvalue of $A^T A$.
- (c) Show that if (λ, x, y) is a solution of the system

$$\begin{cases} \nabla \phi(x, y) = \lambda \nabla \gamma(x, y) \\ \gamma(x, y) = 1 \end{cases}$$

then $\phi(x, y) = \lambda$.

- (d) Conclude that $\|A\|$ is the square root of the largest eigenvalue of $A^T A$. [Remember that a positive matrix—i.e., one of the form $A^T A$ —cannot have negative eigenvalues!]

[All these results generalise to $\mathbf{R}^{n \times n}$ for $n > 2$.]

3. (a) Show that if λ is an eigenvalue of A , then λ^2 is an eigenvalue of A^2 .
- (b) Show that if $A \in \mathbf{R}^{n \times n}$ is diagonalisable, and λ is an eigenvalue of A^2 , then $\pm\sqrt{\lambda}$ is an eigenvalue of A .
- (c) Conclude that if A is a symmetric matrix, then

$$\|A\| = \max \{ |\lambda| \mid \lambda \text{ is an eigenvalue of } A \}.$$

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(d) Calculate the norm of each of the following matrices:

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}, \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$$

(e) Show that the parallelogram law fails for this notion of norm of a matrix—i.e., there does not exist an inner product $\mathbf{R}^{2 \times 2} \times \mathbf{R}^{2 \times 2} \xrightarrow{\langle \cdot, \cdot \rangle} \mathbf{R}$ such that $\|A\| = \sqrt{\langle A, A \rangle}$.