

Self-study problem sheet 2

1. Let $A \in \mathbb{R}^{n \times n}$ be a symmetric matrix. We admit without proof the fact that we can write

$$A = V \Sigma V^T$$

where V is a matrix whose columns are composed of the eigenvectors of A satisfying the orthogonality condition $VV^T = V^TV = I$ and

$$\Sigma = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

is the diagonal matrix with the eigenvalues $\{\lambda_j\}_{j=1}^n$ (which are all real) along the diagonal.

- (a) Show that $\|V\mathbf{x}\|_2 = \|\mathbf{x}\|_2$ for all $\mathbf{x} \in \mathbb{R}^n$.
- (b) Show that the spectral radius satisfies

$$\rho(A) = \|A\|_2.$$

2. Let $\|\cdot\|_M$ be the matrix norm induced by a norm $\|\cdot\|_V$ on \mathbb{R}^n . Let $A \in \mathbb{R}^{n \times n}$ be an invertible matrix.

- (a) Show that $K_M(A) \geq 1$
- (b) Show that $K_M(\alpha A) = K_M(A)$ for all $\alpha \in \mathbb{R}$, $\alpha \neq 0$.
- (c) Show that $K_M(A^{-1}) = K_M(A)$.
- (d) Show that for any *singular* matrix B , we have the bound

$$\frac{\|A - B\|_M}{\|A\|_M} \geq \frac{1}{K_M(A)}.$$

SOLUTIONS

1. (a) We have by definition of the 2-norm $\|V\mathbf{x}\|_2^2 = (V\mathbf{x})^\top V\mathbf{x} = \mathbf{x}^\top V^\top V\mathbf{x} = \mathbf{x}^\top \mathbf{x} = \|\mathbf{x}\|_2^2$.
- (b) Since we have $\rho(A) \leq \|A\|_2$ from the lectures, it is enough to show that $\rho(A) \geq \|A\|_2$.

For any $\mathbf{y} \in \mathbb{R}^n \setminus \{\mathbf{0}\}$, we have $\mathbf{y} = V\mathbf{x}$ for $\mathbf{x} = V^\top \mathbf{y} \neq \mathbf{0}$. Therefore

$$\frac{\|A\mathbf{y}\|_2}{\|\mathbf{y}\|_2} = \frac{\|AV\mathbf{x}\|_2}{\|V\mathbf{x}\|_2} = \frac{\|V\Sigma V^\top V\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = \frac{\|V\Sigma\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = \frac{\|\Sigma\mathbf{x}\|_2}{\|\mathbf{x}\|_2}$$

where we have used $\|V\mathbf{x}\|_2 = \|\mathbf{x}\|_2$ and $\|V\Sigma\mathbf{x}\|_2 = \|\Sigma\mathbf{x}\|_2$ from (a) above.

Therefore, we have

$$\|A\|_2 = \sup_{\mathbf{y} \in \mathbb{R}^n \setminus \{\mathbf{0}\}} \frac{\|A\mathbf{y}\|_2}{\|\mathbf{y}\|_2} = \sup_{\mathbf{x} \in \mathbb{R}^n \setminus \{\mathbf{0}\}} \frac{\|\Sigma\mathbf{x}\|_2}{\|\mathbf{x}\|_2} = \sup_{\mathbf{x} \in \mathbb{R}^n \setminus \{\mathbf{0}\}} \frac{\sqrt{\sum_{i=1}^n |\lambda_i|^2 |x_i|^2}}{\sqrt{\sum_{i=1}^n |x_i|^2}}$$

Clearly $\sum_{i=1}^n |\lambda_i|^2 |x_i|^2 \leq \rho(A)^2 \sum_{i=1}^n |x_i|^2$ since $\rho(A) \geq |\lambda_i|$ for all eigenvalues λ_i of A . From this we then get

$$\|A\|_2 \leq \rho(A)$$

which completes the proof.

2. (a) Since $AA^{-1} = I$ we get $1 = \|I\|_M \leq \|A\|_M \|A^{-1}\|_M = K_M(A)$.
- (b) Since $(\alpha A)^{-1} = \frac{1}{\alpha} A^{-1}$ we get $\|(\alpha A)\|_M \|(\alpha A)^{-1}\|_M = \|A\|_M \|A^{-1}\|_M$. This gives the result.
- (c) This follows from $(A^{-1})^{-1} = A$.
- (d) B is singular implies that there exists a vector $\mathbf{v} \neq 0$ such that $B\mathbf{v} = 0$. We then write $B = A - (A - B)$ to obtain

$$\mathbf{0} = B\mathbf{v} = A\mathbf{v} - (A - B)\mathbf{v} \iff A\mathbf{v} = (A - B)\mathbf{v} \iff \mathbf{v} = (I - A^{-1}B)\mathbf{v}.$$

Therefore, by definition of the induced norm,

$$\|(I - A^{-1}B)\|_M \geq \frac{\|(I - A^{-1}B)\mathbf{v}\|_V}{\|\mathbf{v}\|_V} = \frac{\|\mathbf{v}\|_V}{\|\mathbf{v}\|_V} = 1.$$

But then $\|(I - A^{-1}B)\|_M = \|A^{-1}(A - B)\|_M \leq \|A^{-1}\|_M \|A - B\|_M$, and thus

$$\frac{\|A - B\|_M}{\|A\|_M} \geq \frac{\|(I - A^{-1}B)\|_M}{\|A\|_M \|A^{-1}\|_M} \geq \frac{1}{K_M(A)}.$$