

# Chapter 7

## Norms

In this chapter, the field  $K$  is always  $\mathbb{R}$  or  $\mathbb{C}$  and  $E$  denotes  $K^n$ . The scalar (if  $K = \mathbb{R}$ ) or Hermitian (if  $K = \mathbb{C}$ ) product on  $E$  is denoted by  $\langle x, y \rangle := \sum_j \bar{x}_j y_j$ .

**Definition 7.1** If  $A \in \mathbf{M}_n(K)$ , the spectral radius of  $A$ , denoted by  $\rho(A)$ , is the largest modulus of the eigenvalues of  $A$ :

$$\rho(A) = \max\{|\lambda|; \lambda \in \text{Sp}(A)\}.$$

When  $K = \mathbb{R}$ , this takes into account the complex eigenvalues when computing  $\rho(A)$ .

### 7.1 A Brief Review

#### 7.1.1 The $\ell^p$ Norms

The vector space  $E$  is endowed with various norms, pairwise equivalent because  $E$  has finite dimension (Proposition 7.3 below). Among these, the most used norms are the  $\ell^p$  norms:

$$\|x\|_p = \left( \sum_j |x_j|^p \right)^{1/p}, \quad \|x\|_\infty = \max_j |x_j|.$$

**Proposition 7.1** For  $1 \leq p \leq \infty$ , the map  $x \mapsto \|x\|_p$  is a norm on  $E$ . In particular, one has Minkowski's inequality

$$\|x + y\|_p \leq \|x\|_p + \|y\|_p. \tag{7.1}$$

Furthermore, one has Hölder's inequality

$$|\langle x, y \rangle| \leq \|x\|_p \|y\|_{p'}, \quad \frac{1}{p} + \frac{1}{p'} = 1. \quad (7.2)$$

The numbers  $p, p'$  are called *conjugate exponents*.

*Proof.* Everything is obvious except perhaps the Hölder and Minkowski inequalities. When  $p = 1$  or  $p = \infty$ , these inequalities are trivial. We thus assume that  $1 < p < \infty$ .

Let us begin with (7.2). If  $x$  or  $y$  is null, it is obvious. Indeed, one can even assume, by decreasing the value of  $n$ , that none of the  $x_j, y_j$ s is null. Likewise, because  $|\langle x, y \rangle| \leq \sum_j |x_j| |y_j|$ , one can also assume that the  $x_j, y_j$  are real and positive. Dividing by  $\|x\|_p$  and by  $\|y\|_{p'}$ , one may restrict attention to the case where  $\|x\|_p = \|y\|_{p'} = 1$ . Hence,  $x_j, y_j \in (0, 1]$  for every  $j$ . Let us define

$$a_j = p \log x_j, \quad b_j = p' \log y_j.$$

Because the exponential function is convex,

$$e^{a_j/p + b_j/p'} \leq \frac{1}{p} e^{a_j} + \frac{1}{p'} e^{b_j};$$

that is,

$$x_j y_j \leq \frac{1}{p} x_j^p + \frac{1}{p'} y_j^{p'}.$$

Summing over  $j$ , we obtain

$$\langle x, y \rangle \leq \frac{1}{p} \|x\|_p^p + \frac{1}{p'} \|y\|_{p'}^{p'} = \frac{1}{p} + \frac{1}{p'} = 1,$$

which proves (7.2).

We now turn to (7.1). First, we have

$$\|x + y\|_p^p = \sum_k |x_k + y_k|^p \leq \sum_k |x_k| |x_k + y_k|^{p-1} + \sum_k |y_k| |x_k + y_k|^{p-1}.$$

Let us apply Hölder's inequality to each of the two terms of the right-hand side. For example,

$$\sum_k |x_k| |x_k + y_k|^{p-1} \leq \|x\|_p \left( \sum_k |x_k + y_k|^{(p-1)p'} \right)^{1/p'},$$

which amounts to

$$\sum_k |x_k| |x_k + y_k|^{p-1} \leq \|x\|_p \|x + y\|_p^{p-1}.$$

Finally,

$$\|x + y\|_p^p \leq (\|x\|_p + \|y\|_p) \|x + y\|_p^{p-1},$$

which gives (7.1).  $\square$

For  $p = 2$ , the norm  $\|\cdot\|_2$  is given by an Hermitian form and thus satisfies the Cauchy–Schwarz inequality:

$$|\langle x, y \rangle| \leq \|x\|_2 \|y\|_2.$$

This is a particular case of Hölder’s inequality.

**Proposition 7.2** *For conjugate exponents  $p, p'$ , one has*

$$\|x\|_p = \sup_{y \neq 0} \frac{\Re \langle x, y \rangle}{\|y\|_{p'}} = \sup_{y \neq 0} \frac{|\langle x, y \rangle|}{\|y\|_{p'}}.$$

*Proof.* The inequality  $\geq$  is a consequence of Hölder’s. The reverse inequality is obtained by taking  $y_j = \bar{x}_j |x_j|^{p-2}$  if  $p < \infty$ . If  $p = \infty$ , choose  $y_j = \bar{x}_j$  for an index  $j$  such that  $|x_j| = \|x\|_\infty$ . For  $k \neq j$ , take  $y_k = 0$ .  $\square$

### 7.1.2 Equivalent Norms

**Definition 7.2** *Two norms  $N$  and  $N'$  on a (real or complex) vector space are said to be equivalent if there exist two numbers  $c, c' \in \mathbb{R}$  such that*

$$N \leq cN', \quad N' \leq c'N.$$

The equivalence between norms is obviously an equivalence relation, as its name implies. As announced above, we have the following result.

**Proposition 7.3** *All norms on  $E = K^n$  are equivalent. For example,*

$$\|x\|_\infty \leq \|x\|_p \leq n^{1/p} \|x\|_\infty.$$

*Proof.* It is sufficient to show that every norm is equivalent to  $\|\cdot\|_1$ .

Let  $N$  be a norm on  $E$ . If  $x \in E$ , the triangle inequality gives

$$N(x) \leq \sum_i |x_i| N(\mathbf{e}^i),$$

where  $(\mathbf{e}^1, \dots, \mathbf{e}^n)$  is the canonical basis. One thus has  $N \leq c\|\cdot\|_1$  for  $c := \max_i N(\mathbf{e}^i)$ . Observe that this first inequality expresses the fact that  $N$  is Lipschitz (hence continuous) on the metric space  $X = (E, \|\cdot\|_1)$ .

For the reverse inequality, we reduce *ad absurdum*. Let us assume that the supremum of  $\|x\|_1/N(x)$  is infinite for  $x \neq 0$ . By homogeneity, there would then exist a sequence of vectors  $(x^m)_{m \in \mathbb{N}}$  such that  $\|x^m\|_1 = 1$  and  $N(x^m) \rightarrow 0$  when  $m \rightarrow +\infty$ . The unit sphere of  $X$  is compact, thus one may assume (up to the extraction of a subsequence) that  $x^m$  converges to a vector  $x$  such that  $\|x\|_1 = 1$ . In particular,  $x \neq 0$ . Because  $N$  is continuous on  $X$ , one has also  $N(x) = \lim_{m \rightarrow +\infty} N(x^m) = 0$  and because  $N$  is a norm, we deduce  $x = 0$ , a contradiction.  $\square$

### 7.1.3 Duality

**Definition 7.3** Given a norm  $\|\cdot\|$  on  $K^n$  ( $K = \mathbb{R}$  or  $\mathbb{C}$ ), its dual norm on  $K^n$  is defined by

$$\|x\|' := \sup_{y \neq 0} \frac{\Re \langle x, y \rangle}{\|y\|},$$

or equivalently

$$\|x\|' := \sup_{y \neq 0} \frac{|\langle x, y \rangle|}{\|y\|}.$$

The fact that  $\|\cdot\|'$  is a norm is obvious. For every  $x, y \in K^n$ , one has

$$|\langle x, y \rangle| \leq \|x\| \cdot \|y\|'. \quad (7.3)$$

Proposition 7.2 shows that the dual norm of  $\|\cdot\|_p$  is  $\|\cdot\|_q$  for  $1/p + 1/q = 1$ . Because  $p \mapsto q$  is an involution, this suggests the following property:

**Proposition 7.4** The bi-dual (dual of the dual norm) of a norm is this norm itself:

$$(\|\cdot\|')' = \|\cdot\|.$$

*Proof.* From (7.3), one has  $(\|\cdot\|')' \leq \|\cdot\|$ . The converse is a consequence of the Hahn–Banach theorem: the unit ball  $B$  of  $\|\cdot\|$  is convex and compact. If  $x$  is a point of its boundary (i.e.,  $\|x\| = 1$ ), there exists an  $\mathbb{R}$ -affine (i.e., of the form constant plus  $\mathbb{R}$ -linear) function that vanishes at  $x$  and is nonpositive on  $B$ . Such a function can be written in the form  $z \mapsto \Re \langle z, y \rangle + c$ , where  $c$  is a constant, necessarily equal to  $-\Re \langle x, y \rangle$ . Without loss of generality, one may assume that  $\langle y, x \rangle$  is real and non-negative. Hence

$$\|y\|' = \sup_{\|z\|=1} \Re \langle y, z \rangle = \langle y, x \rangle.$$

One deduces

$$(\|x\|')' \geq \frac{\langle y, x \rangle}{\|y\|'} = 1 = \|x\|.$$

By homogeneity, this is true for every  $x \in \mathbb{C}^n$ .  $\square$

### 7.1.4 Matrix Norms

Let us recall that  $\mathbf{M}_n(K)$  can be identified with the set of endomorphisms of  $E = K^n$  by

$$A \mapsto (x \mapsto Ax).$$

**Definition 7.4** If  $\|\cdot\|$  is a norm on  $E$  and if  $A \in \mathbf{M}_n(K)$ , we define

$$\|A\| := \sup_{x \neq 0} \frac{\|Ax\|}{\|x\|}.$$

Equivalently,

$$\|A\| = \sup_{\|x\| \leq 1} \|Ax\| = \max_{\|x\| \leq 1} \|Ax\|.$$

One verifies easily that  $A \mapsto \|A\|$  is a norm on  $\mathbf{M}_n(K)$ . It is called the *norm induced* by that of  $E$ , or the norm *subordinated* to that of  $E$ . Although we adopted the same notation  $\|\cdot\|$  for the two norms, that on  $E$  and that on  $\mathbf{M}_n(K)$ , these are, of course, distinct objects. In many places, one finds the notation  $\|\cdot\|$  for the induced norm. When one does not wish to mention by which norm on  $E$  a given norm on  $\mathbf{M}_n(K)$  is induced, one says that  $A \mapsto \|A\|$  is a *matrix norm*. The main properties of matrix norms are

$$\|AB\| \leq \|A\| \|B\|, \quad \|I_n\| = 1.$$

These properties are those of any *algebra norm*. In particular, one has  $\|A^k\| \leq \|A\|^k$  for every  $k \in \mathbb{N}$ .

### Examples

Three  $l^p$ -matrix norms can be computed in closed form:

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^{i=n} |a_{ij}|,$$

$$\|A\|_\infty = \max_{1 \leq i \leq n} \sum_{j=1}^{j=n} |a_{ij}|,$$

$$\|A\|_2 = \|A^*A\|_2 = \rho(A^*A)^{1/2}.$$

To prove these formulæ, we begin by proving the inequalities  $\geq$ , selecting a suitable vector  $x$ , and writing  $\|A\|_p \geq \|Ax\|_p / \|x\|_p$ . For  $p = 1$  we choose an index  $j$  such that the maximum in the above formula is achieved. Then we let  $x_j = 1$ , and  $x_k = 0$  otherwise. For  $p = \infty$ , we let  $x_j = \bar{a}_{i_0j} / |a_{i_0j}|$ , where  $i_0$  achieves the maximum in the above formula. For  $p = 2$  we choose an eigenvector of  $A^*A$  associated with an eigenvalue of maximal modulus. We thus obtain three inequalities. The reverse inequalities are direct consequences of the definitions. The similarity of the formulæ for  $\|A\|_1$  and  $\|A\|_\infty$ , as well as the equality  $\|A\|_2 = \|A^*\|_2$  illustrate the general formula

$$\|A^*\|_{p'} = \|A\|_p = \sup_{x \neq 0} \sup_{y \neq 0} \frac{\Re(y^*Ax)}{\|x\|_p \cdot \|y\|_{p'}} = \sup_{x \neq 0} \sup_{y \neq 0} \frac{|(y^*Ax)|}{\|x\|_p \cdot \|y\|_{p'}},$$

where again  $p$  and  $p'$  are conjugate exponents.

We point out that if  $H$  is Hermitian, then  $\|H\|_2 = \rho(H^2)^{1/2} = \rho(H)$ . Therefore the spectral radius is a norm over  $\mathbf{H}_n$ , although it is not over  $\mathbf{M}_n(\mathbb{C})$ . We already mentioned this fact, as a consequence of the Weyl inequalities.

**Proposition 7.5** *For an induced norm, the condition  $\|B\| < 1$  implies that  $I_n - B$  is invertible, with the inverse given by the sum of the series*

$$\sum_{k=0}^{\infty} B^k.$$

*Proof.* The series  $\sum_k B^k$  is normally convergent, because  $\sum_k \|B^k\| \leq \sum_k \|B\|^k$ , where the latter series converges because  $\|B\| < 1$ . Because  $\mathbf{M}_n(K)$  is complete, the series  $\sum_k B^k$  converges. Furthermore,  $(I_n - B) \sum_{k \leq N} B^k = I_n - B^{N+1}$ , which tends to  $I_n$ . The sum of the series is thus the inverse of  $I_n - B$ . One has, moreover,

$$\|(I_n - B)^{-1}\| \leq \sum_k \|B\|^k = \frac{1}{1 - \|B\|}.$$

□

One can also deduce Proposition 7.5 from the following statement.

**Proposition 7.6** *For every induced norm, one has*

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

*Proof.* The case  $K = \mathbb{C}$  is easy, because there exists an eigenvector  $X \in E$  associated with an eigenvalue of modulus  $\rho(A)$ :

$$\rho(A)\|X\| = \|\lambda X\| = \|AX\| \leq \|A\| \|X\|.$$

If  $K = \mathbb{R}$ , one needs a more involved trick.

Let us choose a norm on  $\mathbb{C}^n$  and let us denote by  $N$  the induced norm on  $\mathbf{M}_n(\mathbb{C})$ . We still denote by  $N$  its restriction to  $\mathbf{M}_n(\mathbb{R})$ ; it is a norm. This space has finite dimension, thus any two norms are equivalent: there exists  $C > 0$  such that  $N(B) \leq C\|B\|$  for every  $B$  in  $\mathbf{M}_n(\mathbb{R})$ . Using the result already proved in the complex case, one has for every  $m \in \mathbb{N}$  that

$$\rho(A)^m = \rho(A^m) \leq N(A^m) \leq C\|A^m\| \leq C\|A\|^m.$$

Taking the  $m$ th root and letting  $m$  tend to infinity, and noticing that  $C^{1/m}$  tends to 1, one obtains the announced inequality. □

In general, the equality does not hold. For example, if  $A$  is nilpotent although nonzero, one has  $\rho(A) = 0 < \|A\|$  for every matrix norm.

**Proposition 7.7** *Let  $\|\cdot\|$  be a norm on  $K^n$  and  $P \in \mathbf{GL}_n(K)$ . Hence,  $N(x) := \|Px\|$  defines a norm on  $K^n$ . Denoting still by  $\|\cdot\|$  and  $N$  the induced norms on  $K^n$ , one has  $N(A) = \|PAP^{-1}\|$ .*

*Proof.* Using the change of dummy variable  $y = Px$ , we have

$$N(A) = \sup_{x \neq 0} \frac{\|PAx\|}{\|Px\|} = \sup_{y \neq 0} \frac{\|PAP^{-1}y\|}{\|y\|} = \|PAP^{-1}\|.$$

□

## 7.2 Householder's Theorem

Householder's theorem is a kind of converse of the inequality  $\rho(B) \leq \|B\|$ .

**Theorem 7.1** *For every  $B \in \mathbf{M}_n(\mathbb{C})$  and all  $\varepsilon > 0$ , there exists a norm on  $\mathbb{C}^n$  such that for the induced norm,*

$$\|B\| \leq \rho(B) + \varepsilon.$$

In other words,  $\rho(B)$  is the infimum of  $\|B\|$ , as  $\|\cdot\|$  ranges over the set of matrix norms.

*Proof.* From Theorem 3.5 there exists  $P \in \mathbf{GL}_n(\mathbb{C})$  such that  $T := PBP^{-1}$  is upper-triangular. From Proposition 7.7, one has

$$\inf \|B\| = \inf \|PBP^{-1}\| = \inf \|T\|,$$

where the infimum is taken over the set of induced norms. Because  $B$  and  $T$  have the same spectra, hence the same spectral radius, it is enough to prove the theorem for upper-triangular matrices.

For such a matrix  $T$ , Proposition 7.7 still gives

$$\inf \|T\| \leq \inf \{\|QTQ^{-1}\|_2; Q \in \mathbf{GL}_n(\mathbb{C})\}.$$

Let us now take  $Q(\mu) = \text{diag}(1, \mu, \mu^2, \dots, \mu^{n-1})$ . The matrix  $Q(\mu)TQ(\mu)^{-1}$  is upper-triangular, with the same diagonal as that of  $T$ . Indeed, the entry with indices  $(i, j)$  becomes  $\mu^{i-j}t_{ij}$ . Hence,

$$\lim_{\mu \rightarrow \infty} Q(\mu)TQ(\mu)^{-1}$$

is simply the matrix  $D = \text{diag}(t_{11}, \dots, t_{nn})$ . Because  $\|\cdot\|_2$  is continuous (as is every norm), one deduces

$$\inf \|T\| \leq \lim_{\mu \rightarrow \infty} \|Q(\mu)TQ(\mu)^{-1}\|_2 = \|D\|_2 = \sqrt{\rho(D^*D)} = \max |t_{jj}| = \rho(T).$$

□

**Remark**

The theorem tells us that  $\rho(A) = \Lambda(A)$ , where

$$\Lambda(A) := \inf \|A\|,$$

the infimum being taken over the set of matrix norms. The first part of the proof tells us that  $\rho$  and  $\Lambda$  coincide on the set of diagonalizable matrices, which is a dense subset of  $\mathbf{M}_n(\mathbb{C})$ . But this is insufficient to conclude, since  $\Lambda$  is a priori only upper semicontinuous, as the infimum of continuous functions. The continuity of  $\Lambda$  is actually a consequence of the theorem.

An interesting consequence of Householder's theorem is the following link between matrix norms and the spectral radius.

**Proposition 7.8** *If  $A \in \mathbf{M}_n(k)$  (with  $k = \mathbb{R}$  or  $\mathbb{C}$ ), then*

$$\rho(A) = \lim_{m \rightarrow \infty} \|A^m\|^{1/m}$$

for every matrix norm.

*Proof.* Let  $\|\cdot\|$  be a matrix norm over  $\mathbf{M}_n(k)$ . From Proposition 7.6 and the fact that

$$\text{Sp}(A^m) = \{\lambda^m \mid \lambda \in \text{Sp}A\},$$

we have

$$\rho(A) = \rho(A^m)^{1/m} \leq \|A^m\|^{1/m}.$$

Passing to the limit, we have

$$\rho(A) \leq \liminf_{m \rightarrow +\infty} \|A^m\|^{1/m}. \quad (7.4)$$

Conversely, let  $\varepsilon > 0$  be given. From Theorem 7.1, there exists a matrix norm  $N$  over  $\mathbf{M}_n(\mathbb{C})$  such that  $N(A) < \rho(A) + \varepsilon$ . By finite-dimensionality, the restriction of  $N$  over  $\mathbf{M}_n(k)$  (just in case that  $k = \mathbb{R}$ ) is equivalent to  $\|\cdot\|$ : there exists a constant  $c$  such that  $\|\cdot\| \leq cN$ . If  $m \geq 1$ , we thus have

$$\|A^m\| \leq cN(A^m) \leq cN(A)^m < c(\rho(A) + \varepsilon)^m,$$

whence

$$\|A^m\|^{1/m} \leq c^{1/m}(\rho(A) + \varepsilon).$$

Passing to the limit, we have

$$\limsup_{m \rightarrow +\infty} \|A^m\|^{1/m} \leq \rho(A) + \varepsilon.$$

Letting  $\varepsilon$  tend to zero, there remains

$$\limsup_{m \rightarrow +\infty} \|A^m\|^{1/m} \leq \rho(A).$$



Comparing with (7.4), we see that the sequence  $\|A^m\|^{1/m}$  is convergent and its limit equals  $\rho(A)$ .

□

**Comments**

- In the first edition, we proved Proposition 7.8 by applying the Hadamard formula for the convergence radius of the series  $\sum_{j \geq 0} z^j A^j$ .
- A classical lemma in calculus tells us that the limit of  $\|A^m\|^{1/m}$  is also its infimum over  $m$ , because the sequence  $\|A^m\|$  is submultiplicative.

**7.3 An Interpolation Inequality**

**Theorem 7.2 (case  $K = \mathbb{C}$ )** *Let  $\|\cdot\|_p$  be the norm on  $\mathbf{M}_n(\mathbb{C})$  induced by the norm  $l^p$  on  $\mathbb{C}^n$ . The function*

$$\begin{aligned} 1/p &\mapsto \log \|A\|_p, \\ [0, 1] &\rightarrow \mathbb{R}, \end{aligned}$$

*is convex. In other words, if  $1/r = \theta/p + (1 - \theta)/q$  with  $\theta \in (0, 1)$ , then*

$$\|A\|_r \leq \|A\|_p^\theta \|A\|_q^{1-\theta}.$$

**Remarks**

1. The proof uses the fact that  $K = \mathbb{C}$ . However, the norms induced by the  $\|\cdot\|_p$ s on  $\mathbf{M}_n(\mathbb{R})$  and  $\mathbf{M}_n(\mathbb{C})$  take the same values on real matrices, even although their definitions are different (see Exercise 6). The statement is thus still true in  $\mathbf{M}_n(\mathbb{R})$ .
2. The case  $(p, q, r) = (1, \infty, 2)$  admits a direct proof. See the exercises.
3. The result still holds true in infinite dimension, at the expense of some functional analysis. One can even take different  $L^p$  norms at the source and target spaces. Here is an example.

**Theorem 7.3 (Riesz–Thorin)** *Let  $\Omega$  be an open set in  $\mathbb{R}^D$  and  $\omega$  an open set in  $\mathbb{R}^d$ . Let  $p_0, p_1, q_0, q_1$  be four numbers in  $[1, +\infty]$ . Let  $\theta \in [0, 1]$  and  $p, q$  be defined by*

$$\frac{1}{p} = \frac{1 - \theta}{p_0} + \frac{\theta}{p_1}, \quad \frac{1}{q} = \frac{1 - \theta}{q_0} + \frac{\theta}{q_1}.$$

*Consider a linear operator  $T$  defined on  $L^{p_0} \cap L^{p_1}(\Omega)$ , taking values in  $L^{q_0} \cap L^{q_1}(\omega)$ . Assume that  $T$  can be extended as a continuous operator from  $L^{p_j}(\Omega)$  to  $L^{q_j}(\omega)$ , with norm  $M_j$ ,  $j = 1, 2$  :*

$$M_j := \sup_{f \neq 0} \frac{\|Tf\|_{q_j}}{\|f\|_{p_j}}.$$

Then  $T$  can be extended as a continuous operator from  $L^p(\Omega)$  to  $L^q(\omega)$ , and its norm is bounded above by

$$M_0^{1-\theta} M_1^\theta.$$

4. A fundamental application is the continuity of the Fourier transform from  $L^p(\mathbb{R}^d)$  into its dual  $L^{p'}(\mathbb{R}^d)$  when  $1 \leq p \leq 2$ . We have only to observe that

$$(p_0, p_1, q_0, q_1) = (1, 2, +\infty, 2)$$

is suitable. It can be proved by inspection that every pair  $(p, q)$  such that the Fourier transform is continuous from  $L^p(\mathbb{R}^d)$  into  $L^q(\mathbb{R}^d)$  has the form  $(p, p')$  with  $1 \leq p \leq 2$ .

5. One has analogous results for the Fourier series. Therein lies the origin of the Riesz–Thorin theorem.

*Proof.* (Due to Riesz)

Let us fix  $x$  and  $y$  in  $K^n$ . We have to bound

$$|\langle y, Ax \rangle| = \left| \sum_{j,k} a_{jk} x_j \bar{y}_k \right|.$$

Let  $B$  be the strip in the complex plane defined by  $\Re z \in [0, 1]$ . Given  $z \in B$ , define “conjugate” exponents  $r(z)$  and  $r'(z)$  by

$$\frac{1}{r(z)} = \frac{z}{p} + \frac{1-z}{q}, \quad \frac{1}{r'(z)} = \frac{z}{p'} + \frac{1-z}{q'}.$$

Set

$$\begin{aligned} X_j(z) &:= |x_j|^{-1+r/r(z)} x_j = x_j \exp\left(\left(\frac{r}{r(z)} - 1\right) \log |x_j|\right), \\ Y_j(z) &:= |y_j|^{-1+r'/r'(z)} y_j. \end{aligned}$$

We then have

$$\|X(z)\|_{r(\Re z)} = \|x\|_r^{r/r(\Re z)}, \quad \|Y(z)\|_{r'(\Re z)} = \|y\|_{r'}^{r'/r'(\Re z)}.$$

Next, define a holomorphic map in the strip  $B$  by  $f(z) := \langle Y(z), AX(z) \rangle$ . It is bounded, because the numbers  $X_j(z)$  and  $Y_k(z)$  are. For example,

$$|X_j(z)| = |x_j|^{r/r(\Re z)}$$

lies between  $|x_j|^{r/p}$  and  $|x_j|^{r/q}$ .

Let us set  $M(\theta) = \sup\{|f(z)|; \Re z = \theta\}$ . Hadamard's *three-line lemma* (see [33], Chapter 12, Exercise 8) tells us that,  $f$  being bounded and holomorphic in the strip,

$$\theta \mapsto \log M(\theta)$$

is convex on  $(0, 1)$ . However,  $r(0) = q$ ,  $r(1) = p$ ,  $r'(0) = q'$ ,  $r'(1) = p'$ ,  $r(\theta) = r$ ,  $r'(\theta) = r'$ ,  $X(\theta) = x$ , and  $Y(\theta) = y$ . Hence

$$|\langle y, Ax \rangle| = |f(\theta)| \leq M(\theta) \leq M(1)^\theta M(0)^{1-\theta}.$$

Now we have

$$\begin{aligned} M(1) &= \sup\{|f(z)|; \Re z = 1\} \\ &\leq \sup\{\|AX(z)\|_{r(1)} \|Y(z)\|_{r(1)'}; \Re z = 1\} \\ &= \sup\{\|AX(z)\|_p \|Y(z)\|_{p'}; \Re z = 1\} \\ &\leq \|A\|_p \sup\{\|X(z)\|_p \|Y(z)\|_{p'}; \Re z = 1\} \\ &= \|A\|_p \|x\|_r^{r/p} \|y\|_{r'}^{r'/p'}. \end{aligned}$$

Likewise,  $M(0) \leq \|A\|_q \|x\|_r^{r/q} \|y\|_{r'}^{r'/q'}$ . Hence

$$\begin{aligned} |\langle y, Ax \rangle| &\leq \|A\|_p^\theta \|A\|_q^{1-\theta} \|x\|_r^{r(\theta/p + (1-\theta)/q)} \|y\|_{r'}^{r'(\theta/p' + (1-\theta)/q')} \\ &= \|A\|_p^\theta \|A\|_q^{1-\theta} \|x\|_r \|y\|_{r'}. \end{aligned}$$

Finally,

$$\|Ax\|_r = \sup_{y \neq 0} \frac{|\langle y, Ax \rangle|}{\|y\|_{r'}} \leq \|A\|_p^\theta \|A\|_q^{1-\theta} \|x\|_r,$$

which proves the theorem.  $\square$

## 7.4 Von Neumann's Inequality

We say that a matrix  $M \in \mathbf{M}_n(\mathbb{C})$  is a *contraction* if  $\|Mx\|_2 \leq \|x\|_2$  for every vector  $x$ , or equivalently  $\|M\|_2 \leq 1$ . Developing in the form  $x^* M^* M x \leq x^* x$ , this translates as  $M^* M \leq I_n$  in the sense of Hermitian matrices. Inasmuch as  $\|M^*\|_2 = \|M\|_2$ , the Hermitian conjugate of a contraction is a contraction. When  $U$  and  $V$  are unitary matrices,  $U^* M V$  is a contraction if and only if  $M$  is a contraction.

The following statement is due to von Neumann.

**Theorem 7.4** *Let  $M \in \mathbf{M}_n(\mathbb{C})$  be a contraction and  $P \in \mathbb{C}[X]$  be a polynomial. Then*

$$\|P(M)\|_2 \leq \sup\{|P(z)|; z \in \mathbb{C}, |z| = 1\}.$$



We now choose  $k \geq (d^o P)/2$ , so that  $P(M)$  is a block of  $P(V_k)$ . We have  $P(M) = \Pi P(V_k) \Pi^*$  where

$$\Pi = (\dots, 0_n, I_n, 0_n, \dots).$$

If  $x \in \mathbb{C}^n$ , let  $X \in \mathbb{C}^{\ell n}$  denote the vector  $\Pi^* x$ . We have  $\|X\|_2 = \|x\|_2$ . Because  $\Pi$  is an orthogonal projection, we find

$$\begin{aligned} \|P(M)x\|_2 &= \|\Pi P(V_k)X\|_2 \leq \|P(V_k)X\|_2 \leq \sup\{|P(z)| \mid z \in \mathbb{C}, |z| = 1\} \|X\|_2 \\ &= \sup\{|P(z)| \mid z \in \mathbb{C}, |z| = 1\} \|x\|_2, \end{aligned}$$

where we have used (7.5). This is exactly

$$\|P(M)\|_2 \leq \sup\{|P(z)| \mid z \in \mathbb{C}, |z| = 1\}.$$

□

### Exercises

1. Under what conditions on the vectors  $a, b \in \mathbb{C}^n$  does the matrix  $M$  defined by  $m_{ij} = a_i b_j$  satisfy  $\|M\|_p = 1$  for every  $p \in [1, \infty]$ ?
2. Under what conditions on  $x, y$ , and  $p$  does the equality in (7.2) or (7.1) hold?
3. Show that

$$\lim_{p \rightarrow +\infty} \|x\|_p = \|x\|_\infty, \quad \forall x \in E.$$

4. A norm on  $K^n$  is a *strictly convex* norm if  $\|x\| = \|y\| = 1, x \neq y$ , and  $0 < \theta < 1$  imply  $\|\theta x + (1 - \theta)y\| < 1$ .
  - a. Show that  $\|\cdot\|_p$  is strictly convex for  $1 < p < \infty$ , but is not so for  $p = 1, \infty$ .
  - b. Deduce from Corollary 8.1 that the induced norm  $\|\cdot\|_p$  is not strictly convex on  $\mathbf{M}_n(\mathbb{R})$ .
5. Let  $N$  be a norm on  $\mathbb{R}^n$ .
  - a. For  $x \in \mathbb{C}^n$ , define

$$N_1(x) := \inf \left\{ \sum_{\ell} |\alpha_{\ell}| N(x^{\ell}) \right\},$$

where the infimum is taken over the set of decompositions  $x = \sum_{\ell} \alpha_{\ell} x^{\ell}$  with  $\alpha_{\ell} \in \mathbb{C}$  and  $x^{\ell} \in \mathbb{R}^n$ . Show that  $N_1$  is a norm on  $\mathbb{C}^n$  (as a  $\mathbb{C}$ -vector space) whose restriction to  $\mathbb{R}^n$  is  $N$ . **Note:**  $N_1$  is called the *complexification* of  $N$ .

- b. Same question as above for  $N_2$ , defined by

$$N_2(x) := \frac{1}{2\pi} \int_0^{2\pi} [e^{i\theta} x] d\theta,$$

where

$$[x] := \sqrt{N(\Re x)^2 + N(\Im x)^2}.$$

- c. Show that  $N_2 \leq N_1$ .  
 d. If  $N(x) = \|x\|_1$ , show that  $N_1(x) = \|x\|_1$ . Considering then the vector

$$x = \begin{pmatrix} 1 \\ i \end{pmatrix},$$

show that  $N_2 \neq N_1$ .

6. (Continuation of Exercise 5)

The norms  $N$  (on  $\mathbb{R}^n$ ) and  $N_1$  (on  $\mathbb{C}^n$ ) lead to induced norms on  $\mathbf{M}_n(\mathbb{R})$  and  $\mathbf{M}_n(\mathbb{C})$ , respectively. Show that if  $M \in \mathbf{M}_n(\mathbb{R})$ , then  $N(M) = N_1(M)$ . Deduce that Theorem 7.2 holds true in  $\mathbf{M}_n(\mathbb{R})$ .

7. Let  $\|\cdot\|$  be an algebra norm on  $\mathbf{M}_n(K)$  ( $K = \mathbb{R}$  or  $\mathbb{C}$ ), that is, a norm satisfying  $\|AB\| \leq \|A\| \cdot \|B\|$ . Show that  $\rho(A) \leq \|A\|$  for every  $A \in \mathbf{M}_n(K)$ .

8. Let  $B \in \mathbf{M}_n(\mathbb{C})$  be given. Assume that there exists an induced norm such that  $\|B\| = \rho(B)$ . Let  $\lambda$  be an eigenvalue of maximal modulus and  $X$  a corresponding eigenvector. Show that  $X$  does not belong to the range of  $B - \lambda I_n$ . Deduce that the Jordan block associated with  $\lambda$  is diagonal (Jordan reduction is presented in Chapter 9).

9. (Continuation of Exercise 8)

Conversely, show that if the Jordan blocks of  $B$  associated with the eigenvalues of maximal modulus of  $B$  are diagonal, then there exists a norm on  $\mathbb{C}^n$  such that, using the induced norm,  $\rho(B) = \|B\|$ .

10. Here is another proof of Theorem 7.1. Let  $K = \mathbb{R}$  or  $\mathbb{C}$ ,  $A \in \mathbf{M}_n(K)$ , and let  $N$  be a norm on  $K^n$ . If  $\varepsilon > 0$ , we define for all  $x \in K^n$

$$\|x\| := \sum_{k \in \mathbb{N}} (\rho(A) + \varepsilon)^{-k} N(A^k x).$$

- a. Show that this series is convergent (use Proposition 7.8).  
 b. Show that  $\|\cdot\|$  is a norm on  $K^n$ .  
 c. Show that for the induced norm,  $\|A\| \leq \rho(A) + \varepsilon$ .

11. A norm  $\|\cdot\|$  on  $\mathbf{M}_n(\mathbb{C})$  is said to be *unitarily invariant* if  $\|UAV\| = \|A\|$  for every  $A \in \mathbf{M}_n(\mathbb{C})$  and all unitary matrices  $U, V$ .

- a. Find, among the most classical norms, two examples of unitarily invariant norms.  
 b. Given a unitarily invariant norm, show that there exists a norm  $N$  on  $\mathbb{R}^n$  such that

$$\|A\| = N(s_1(A), \dots, s_n(A)),$$

where the  $s_j(A)$ s, the eigenvalues of  $H$  in the polar decomposition  $A = QH$  (see Section 11.4 for this notion), are called the *singular values* of  $A$ .

12. (Bhatia [5]) Suppose we are given a norm  $\|\cdot\|$  on  $\mathbf{M}_n(\mathbb{C})$  that is unitarily invariant (see the previous exercise). If  $A \in \mathbf{M}_n(\mathbb{C})$ , we denote by  $D(A)$  the diagonal matrix obtained by keeping only the  $a_{jj}$  and setting all the other entries to zero. If  $\sigma$  is a permutation, we denote by  $A^\sigma$  the matrix whose entry of index  $(j, k)$  equals  $a_{jk}$  if  $k = \sigma(j)$ , and zero otherwise. For example,  $A^{id} = D(A)$ , where  $id$  is the identity permutation. If  $r$  is an integer between  $1 - n$  and  $n - 1$ , we denote by  $D_r(A)$  the matrix whose entry of index  $(j, k)$  equals  $a_{jk}$  if  $k - j = r$ , and zero otherwise. For example,  $D_0(A) = D(A)$ .

- a. Let  $\omega = \exp(2i\pi/n)$  and let  $U$  be the diagonal matrix whose diagonal entries are the roots of unity  $1, \omega, \dots, \omega^{n-1}$ . Show that

$$D(A) = \frac{1}{n} \sum_{j=0}^{n-1} U^{*j} A U^j.$$

Deduce that  $\|D(A)\| \leq \|A\|$ .

- b. Show that  $\|A^\sigma\| \leq \|A\|$  for every  $\sigma \in S_n$ . Observe that  $\|P\| = \|I_n\|$  for every permutation matrix  $P$ . Show that  $\|M\| \leq \|I_n\|$  for every bistochastic matrix  $M$  (see Section 8.5 for this notion).  
 c. If  $\theta \in \mathbb{R}$ , let us denote by  $U_\theta$  the diagonal matrix, whose  $k$ th diagonal term equals  $\exp(ik\theta)$ . Show that

$$D_r(A) = \frac{1}{2\pi} \int_0^{2\pi} e^{ir\theta} U_\theta A U_\theta^* d\theta.$$

- d. Deduce that  $\|D_r(A)\| \leq \|A\|$ .  
 e. Let  $p$  be an integer between zero and  $n - 1$ , and set  $r = 2p + 1$ . Let us denote by  $T_r(A)$  the matrix whose entry of index  $(j, k)$  equals  $a_{jk}$  if  $|k - j| \leq p$ , and zero otherwise. For example,  $T_3(A)$  is a tridiagonal matrix. Show that

$$T_r(A) = \frac{1}{2\pi} \int_0^{2\pi} d_p(\theta) U_\theta A U_\theta^* d\theta,$$

where

$$d_p(\theta) = \sum_{-p}^p e^{ik\theta}$$

is the *Dirichlet kernel*.

- f. Deduce that  $\|T_r(A)\| \leq L_p \|A\|$ , where

$$L_p = \frac{1}{2\pi} \int_0^{2\pi} |d_p(\theta)| d\theta$$

is the *Lebesgue constant* (**Note:**  $L_p = 4\pi^{-2} \log p + O(1)$ ).

- g. Let  $\Delta(A)$  be the upper-triangular matrix whose entries above the diagonal coincide with those of  $A$ . Using the matrix

$$B = \begin{pmatrix} 0_n & \Delta(A)^* \\ \Delta(A) & 0_n \end{pmatrix},$$

show that  $\|\Delta(A)\|_2 \leq L_n \|A\|_2$  (observe that  $\|B\|_2 = \|\Delta(A)\|_2$ ).

h. What inequality do we obtain for  $\Delta_0(A)$ , the strictly upper-triangular matrix whose entries lying strictly above the diagonal coincide with those of  $A$ ?

13. We endow  $\mathbb{C}^n$  with the usual Hermitian structure, so that  $\mathbf{M}_n(\mathbb{C})$  is equipped with the norm  $\|A\|_2 = \rho(A^*A)^{1/2}$ .

Suppose we are given a sequence of matrices  $(A_j)_{j \in \mathbb{Z}}$  in  $\mathbf{M}_n(\mathbb{C})$  and a summable sequence  $\gamma \in l^1(\mathbb{Z})$  of positive real numbers. Assume, finally, that for every pair  $(j, k) \in \mathbb{Z} \times \mathbb{Z}$ ,

$$\|A_j^* A_k\|_2 \leq \gamma(j-k)^2, \quad \|A_j A_k^*\|_2 \leq \gamma(j-k)^2.$$

a. Let  $F$  be a finite subset of  $\mathbb{Z}$ . Let  $B_F$  denote the sum of the  $A_j$ s as  $j$  runs over  $F$ . Show that

$$\|(B_F^* B_F)^{2m}\|_2 \leq \text{card } F \|\gamma\|_1^{2m}, \quad \forall m \in \mathbb{N}.$$

b. Deduce that  $\|B_F\|_2 \leq \|\gamma\|_1$ .

c. Show (*Cotlar's lemma*) that for every  $x, y \in \mathbb{C}^n$ , the series

$$y^T \sum_{j \in \mathbb{Z}} A_j x$$

is convergent, and that its sum  $y^T A x$  defines a matrix  $A \in \mathbf{M}_n(\mathbb{C})$  that satisfies

$$\|A\| \leq \sum_{j \in \mathbb{Z}} \gamma(j).$$

**Hint:** For a sequence  $(u_j)_{j \in \mathbb{Z}}$  of real numbers, the series  $\sum_j u_j$  is absolutely convergent if and only if there exists  $M < +\infty$  such that  $\sum_{j \in F} |u_j| \leq M$  for every finite subset  $F$ .

d. Deduce that the series  $\sum_j A_j$  converges in  $\mathbf{M}_n(\mathbb{C})$ . May one conclude that it converges normally?

14. Let  $\|\cdot\|$  be an induced norm on  $\mathbf{M}_n(\mathbb{R})$ . We wish to characterize the matrices  $B \in \mathbf{M}_n(\mathbb{R})$  such that there exist  $\varepsilon_0 > 0$  and  $\omega > 0$  with

$$(0 < \varepsilon < \varepsilon_0) \implies (\|I_n - \varepsilon B\| \leq 1 - \omega \varepsilon).$$

a. For the norm  $\|\cdot\|_\infty$ , it is equivalent that  $B$  be strictly diagonally dominant.

b. What is the characterization for the norm  $\|\cdot\|_1$ ?

c. For the norm  $\|\cdot\|_2$ , it is equivalent that  $B^T + B$  is positive-definite.

15. Let  $B \in \mathbf{M}_n(\mathbb{C})$  be given.



- a. Returning to the proof of Theorem 7.1, show that for every  $\varepsilon > 0$  there exists on  $\mathbb{C}^n$  an Hermitian norm  $\|\cdot\|$  such that for the induced norm  $\|B\| \leq \rho(B) + \varepsilon$ .
  - b. Deduce that  $\rho(B) < 1$  holds if and only if there exists a matrix  $A \in \mathbf{HPD}_n$  such that  $A - B^*AB \in \mathbf{HPD}_n$ .
16. Let  $A \in \mathbf{M}_n(\mathbb{C})$  be a diagonalizable matrix:  $A = S \operatorname{diag}(d_1, \dots, d_n) S^{-1}$ . Let  $\|\cdot\|$  be an induced norm for which  $\|D\| = \max_j |d_j|$  holds, where

$$D := \operatorname{diag}(d_1, \dots, d_n).$$

Show that for every  $E \in \mathbf{M}_n(\mathbb{C})$  and for every eigenvalue  $\lambda$  of  $A + E$ , there exists an index  $j$  such that

$$|\lambda - d_j| \leq \|S\| \cdot \|S^{-1}\| \cdot \|E\|.$$

17. Let  $A \in \mathbf{M}_n(K)$ , with  $K = \mathbb{R}$  or  $\mathbb{C}$ . Give another proof, using the Cauchy-Schwarz inequality, of the following particular case of Theorem 7.2:

$$\|A\|_2 \leq \|A\|_1^{1/2} \|A\|_\infty^{1/2}.$$

18. Show that if  $A \in \mathbf{M}_n(\mathbb{C})$  is normal, then  $\rho(A) = \|A\|_2$ . Deduce that if  $A$  and  $B$  are normal,  $\rho(AB) \leq \rho(A)\rho(B)$ .
19. Let  $N_1$  and  $N_2$  be two norms on  $\mathbb{C}^n$ . Denote by  $\mathcal{N}_1$  and  $\mathcal{N}_2$  the induced norms on  $\mathbf{M}_n(\mathbb{C})$ . Let us define

$$R := \max_{x \neq 0} \frac{N_1(x)}{N_2(x)}, \quad S := \max_{x \neq 0} \frac{N_2(x)}{N_1(x)}.$$

- a. Show that

$$\max_{A \neq 0} \frac{\mathcal{N}_1(A)}{\mathcal{N}_2(A)} = RS = \max_{A \neq 0} \frac{\mathcal{N}_2(A)}{\mathcal{N}_1(A)}.$$

- b. Deduce that if  $\mathcal{N}_1 = \mathcal{N}_2$ , then  $N_2/N_1$  is constant.
- c. Show that if  $\mathcal{N}_1 \leq \mathcal{N}_2$ , then  $N_2/N_1$  is constant and therefore  $\mathcal{N}_2 = \mathcal{N}_1$ .

20. (Continuation of Exercise 19)

Let  $\|\cdot\|$  be an algebra norm on  $\mathbf{M}_n(\mathbb{C})$ . If  $y \in \mathbb{C}^n$  is nonzero, we define  $\|x\|_y := \|\|xy^*\|$ .

- a. Show that  $\|\cdot\|_y$  is a norm on  $\mathbb{C}^n$  for every  $y \neq 0$ .
- b. Let  $\mathcal{N}_y$  be the norm induced by  $\|\cdot\|_y$ . Show that  $\mathcal{N}_y \leq \|\cdot\|$ .
- c. We say that  $\|\cdot\|$  is *minimal* if there exists no other algebra norm less than or equal to  $\|\cdot\|$ . Show that the following assertions are equivalent.
  - i.  $\|\cdot\|$  is an induced norm on  $\mathbf{M}_n(\mathbb{C})$ .
  - ii.  $\|\cdot\|$  is a minimal norm on  $\mathbf{M}_n(\mathbb{C})$ .
  - iii. For all  $y \neq 0$ , one has  $\|\cdot\| = \mathcal{N}_y$ .

21. (Continuation of Exercise 20)

Let  $\|\cdot\|$  be an induced norm on  $\mathbf{M}_n(\mathbb{C})$ .

- Let  $y, z \neq 0$  be two vectors in  $\mathbb{C}^n$ . Show that (with the notation of the previous exercise)  $\|\cdot\|_y / \|\cdot\|_z$  is constant.
- Prove the equality

$$\|xy^*\| \cdot \|zt^*\| = \|xt^*\| \cdot \|zy^*\|.$$

22. Let  $M \in \mathbf{M}_n(\mathbb{C})$  and  $H \in \mathbf{HPD}_n$  be given. Show that

$$\|HMH\|_2 \leq \frac{1}{2} \|H^2M + MH^2\|_2.$$

23. We endow  $\mathbb{R}^2$  with the Euclidean norm  $\|\cdot\|_2$ , and  $\mathbf{M}_2(\mathbb{R})$  with the induced norm, also denoted by  $\|\cdot\|_2$ . We denote by  $\Sigma$  the unit sphere of  $\mathbf{M}_2(\mathbb{R})$ :  $M \in \Sigma$  is equivalent to  $\|M\|_2 = 1$ , that is, to  $\rho(M^T M) = 1$ . Likewise,  $B$  denotes the unit ball of  $\mathbf{M}_2(\mathbb{R})$ .

Recall that if  $C$  is a convex set and if  $P \in C$ , then  $P$  is called an *extremal* point if  $P \in [Q, R]$  and  $Q, R \in C$  imply either  $Q = P$  or  $R = P$ .

- Show that the set of extremal points of  $B$  is equal to  $\mathbf{O}_2(\mathbb{R})$ .
- Show that  $M \in \Sigma$  if and only if there exist two matrices  $P, Q \in \mathbf{O}_2(\mathbb{R})$  and a number  $a \in [0, 1]$  such that

$$M = P \begin{pmatrix} a & 0 \\ 0 & 1 \end{pmatrix} Q.$$

- We denote by  $\mathcal{R} = \mathbf{SO}_2(\mathbb{R})$  the set of rotation matrices, and by  $\mathcal{S}$  that of matrices of planar symmetry. Recall that  $\mathbf{O}_2(\mathbb{R})$  is the disjoint union of  $\mathcal{R}$  and  $\mathcal{S}$ . Show that  $\Sigma$  is the union of the segments  $[r, s]$  as  $r$  runs over  $\mathcal{R}$  and  $s$  runs over  $\mathcal{S}$ .
- Show that two such “open” segments  $(r, s)$  and  $(r', s')$  are either disjoint or equal.
- Let  $M, N \in \Sigma$ . Show that  $\|M - N\|_2 = 2$  (i.e.,  $(M, N)$  is a diameter of  $B$ ) if and only if there exists a segment  $[r, s]$  ( $r \in \mathcal{R}$  and  $s \in \mathcal{S}$ ) such that  $M \in [r, s]$  and  $N \in [-r, -s]$ .

24. (The Banach–Mazur distance)

- Let  $N$  and  $N'$  be two norms on  $k^n$  ( $k = \mathbb{R}$  or  $\mathbb{C}$ ). If  $A \in \mathbf{GL}_n(k)$ , we may define norms

$$\|A\|_{\rightarrow} := \sup_{x \neq 0} \frac{N'(Ax)}{N(x)}, \quad \|A^{-1}\|_{\leftarrow} := \sup_{x \neq 0} \frac{N(A^{-1}x)}{N'(x)}.$$

Show that  $A \mapsto \|A\|_{\rightarrow} \|A^{-1}\|_{\leftarrow}$  achieves its upper bound. We denote by  $\delta(N, N')$  the minimum value. Verify

$$0 \leq \log \delta(N, N'') \leq \log \delta(N, N') + \log \delta(N', N'').$$

When  $N = \|\cdot\|_p$ , we write  $\ell^p$  instead. If in addition  $N' = \|\cdot\|_q$ , we write  $\|\cdot\|_{p,q}$  for  $\|\cdot\|_{\rightarrow}$ .

- b. In the set  $\mathcal{N}$  of norms on  $k^n$ , let us consider the following equivalence relation:  $N \sim N'$  if and only if there exists an  $A \in \mathbf{GL}_n(k)$  such that  $N' = N \circ A$ . Show that  $\log \delta$  induces a metric  $d$  on the quotient set  $\mathbf{Norm} := \mathcal{N} / \sim$ . This metric is called the *Banach–Mazur distance*. How many classes of Hermitian norms are there?
- c. Compute  $\|I_n\|_{p,q}$  for  $1 \leq p, q \leq n$  (there are two cases, depending on the sign of  $q - p$ ). Deduce that

$$\delta(\ell^p, \ell^q) \leq n^\kappa, \quad \kappa := \left| \frac{1}{p} - \frac{1}{q} \right|.$$

- d. Show that  $\delta(\ell^p, \ell^q) = \delta(\ell^{p'}, \ell^{q'})$ , where  $p', q'$  are the conjugate exponents.
- e. i. When  $H \in \mathbf{H}_n$  is positive-semidefinite, find that the average of  $x^* H x$ , as  $x$  runs over the set defined by  $|x_j| = 1$  for all  $j$ s, is  $\text{Tr } H$  (the measure is the product of  $n$  copies of the normalized Lebesgue measure on the unit disk). Deduce that

$$\sqrt{\text{Tr } M^* M} \leq \|M\|_{\infty, 2} := \sup_{x \neq 0} \frac{\|Mx\|_2}{\|x\|_\infty}$$

for every  $M \in \mathbf{M}_n(k)$ .

- ii. Prove also that

$$\|A\|_{p, \infty} = \max_{1 \leq i \leq n} \|A^{(i)}\|_{p'},$$

where  $A^{(i)}$  denotes the  $i$ th row vector of  $A$ .

- iii. Deduce that  $\delta(\ell^2, \ell^\infty) = \sqrt{n}$ .
- iv. Using the triangle inequality for  $\log \delta$ , deduce that

$$\delta(\ell^p, \ell^q) = n^\kappa$$

whenever  $p, q \geq 2$ , and then for every  $p, q$  such that  $(p - 2)(q - 2) \geq 0$ . **Note:** The exact value of  $\delta(\ell^p, \ell^q)$  is not known when  $(p - 2)(q - 2) < 0$ .

- v. Remark that the “curves”  $\{\ell^p \mid 2 \leq p \leq \infty\}$  and  $\{\ell^p \mid 1 \leq p \leq 2\}$  are geodesics, in the sense that the restrictions of the Banach–Mazur distance to these curves satisfy the triangular *equality*.
- f. When  $n = 2$ , prove that  $\delta(\ell^1, \ell^\infty) = 1$ . On the contrary, if  $n \geq 3$ , then prove  $\delta(\ell^1, \ell^\infty) > 1$ .
- g. A theorem proven by John states that the diameter of  $(\mathbf{Norm}, d)$  is precisely  $\frac{1}{2} \log n$ . Show that this metric space is compact. **Note:** One may consider the norm whose unit ball is an  $m$ -agon in  $\mathbb{R}^2$ , with  $m$  even. Denote its class by  $N_m$ . It seems that  $d(\ell^1, N_m) = \frac{1}{2} \log 2$  when  $8 \mid m$ .

25. Given three matrices  $A \in \mathbf{M}_{p \times q}(k)$ ,  $B \in \mathbf{M}_{p \times s}(k)$ , and  $C \in \mathbf{M}_{r \times q}(k)$ , we consider the affine set  $\mathscr{W}$  of matrices  $W \in \mathbf{M}_{n \times m}(k)$  of the form

$$W = \begin{pmatrix} A & B \\ C & D \end{pmatrix},$$

where  $D$  runs over  $\mathbf{M}_{r \times s}(k)$ . Thus  $n = p + r$  and  $m = q + s$ .

Denoting

$$P = \begin{pmatrix} I & \\ & 0 \end{pmatrix}, \quad Q = (I \ 0)$$

the projection matrices, we are going to prove (Parrott's lemma) that

$$\min\{\|W\|_2 \mid W \in \mathscr{W}\} = \max\{\|QW\|_2, \|WP\|_2\}, \quad (7.6)$$

where the right-hand side does not depend on  $D$ :

$$WP = \begin{pmatrix} A \\ C \end{pmatrix}, \quad QW = (A \ B)$$

- a. Check the inequality

$$\inf\{\|W\|_2 \mid W \in \mathscr{W}\} \geq \max\{\|QW\|_2, \|WP\|_2\}.$$

- b. Denote  $\mu(D) := \|W\|_2$ . Show that the infimum of  $\mu$  on  $\mathscr{W}$  is attained.  
 c. Show that it is sufficient to prove (7.6) when  $s = 1$ .  
 d. From now on, we assume that  $s = 1$ , and we consider a matrix  $D_0 \in \mathbf{M}_{r \times 1}(k)$  such that  $\mu$  is minimal at  $D_0$ . We denote by  $W_0$  the associated matrix. Let us introduce a function  $D \mapsto \eta(D) = \mu(D)^2$ . Recall that  $\eta$  is the largest eigenvalue of  $W^*W$ . We denote  $f_0$  its multiplicity when  $D = D_0$ .  
 i. If  $f_0 \geq 2$ , show that  $W_0^*W_0$  has an eigenvector  $v$  with  $v_m = 0$ . Deduce that  $\mu(D_0) \leq \|WP\|_2$ . Conclude in this case.  
 ii. From now on, we suppose  $f_0 = 1$ . Show that  $\eta(D)$  is a simple eigenvalue for every  $D$  in a small neighbourhood of  $D_0$ . Show that  $D \mapsto \eta(D)$  is differentiable at  $D_0$ , and that its differential is given by

$$\Delta \mapsto \frac{2}{\|y\|_2^2} \Re[(QW_0y)^* \Delta Qy],$$

where  $y$  is an associated eigenvector:

$$W_0^*W_0y = \eta(D_0)y.$$

- iii. Deduce that either  $Qy = 0$  or  $QW_0y = 0$ .  
 iv. In the case where  $Qy = 0$ , show that  $\mu(D_0) \leq \|WP\|_2$  and conclude.  
 v. In the case where  $QW_0y = 0$ , prove that  $\mu(D_0) \leq \|QW\|_2$  and conclude.

26. Let  $k$  be  $\mathbb{R}$  or  $\mathbb{C}$ . Given a bounded subset  $F$  of  $\mathbf{M}_n(k)$ , let us denote by  $F_k$  the set of all possible products of  $k$  elements in  $F$ . Given a matrix norm  $\|\cdot\|$ , we denote  $\|F_k\|$  the supremum of the norms of elements of  $F_k$ .

- a. Show that  $\|F_{k+l}\| \leq \|F_k\| \cdot \|F_l\|$ .
- b. Deduce that the sequence  $\|F_k\|^{1/k}$  converges, and that its limit is the infimum of the sequence.
- c. Prove that this limit does not depend on the choice of the matrix norm. This limit is called the *joint spectral radius* of the family  $F$ , and denoted  $\rho(F)$ . This notion is due to Rota and Strang.
- d. Let  $\hat{\rho}(F)$  denote the infimum of  $\|F\|$  when  $\|\cdot\|$  runs over all matrix norms. Show that  $\rho(F) \leq \hat{\rho}(F)$ .
- e. Given a norm  $N$  on  $k^n$  and a number  $\varepsilon > 0$ , we define for every  $x \in k^n$

$$\|x\| := \sum_{l=0}^{\infty} (\rho(F) + \varepsilon)^{-l} \max\{N(Bx) \mid B \in F_l\}.$$

- i. Show that the series converges, and that it defines a norm on  $k^n$ .
  - ii. For the matrix norm associated with  $\|\cdot\|$ , show that  $\|A\| \leq \rho(F) + \varepsilon$  for every  $A \in F$ .
  - iii. Deduce that actually  $\rho(F) = \hat{\rho}(F)$ . Compare with Householder's theorem.
27. (Rota & Strang.) Let  $k$  be  $\mathbb{R}$  or  $\mathbb{C}$ . Given a subset  $F$  of  $\mathbf{M}_n(k)$ , we consider the semi-group  $\mathcal{F}$  generated by  $F$ . It is the union of sets  $F_k$  defined in the previous exercise, as  $k$  runs over  $\mathbb{N}$ . We have  $F_0 = \{I_n\}$ ,  $F_1 = F$ ,  $F_2 = F \cdot F, \dots$

If  $\mathcal{F}$  is bounded, prove that there exists a matrix norm  $\|\cdot\|$  such that  $\|A\| \leq 1$  for every  $A \in F$ . **Hint:** In the previous exercise, take a sup instead of a series.

28. Let  $A \in \mathbf{M}_n(\mathbb{C})$  be given. Let  $\sigma(A)$  be the spectrum of  $A$  and  $\rho(A)$  its complement (the *resolvent set*). For  $\varepsilon > 0$ , we define the  $\varepsilon$ -pseudospectrum of  $A$  as

$$\sigma_\varepsilon(A) := \sigma(A) \cup \left\{ z \in \rho(A); \|(z - A)^{-1}\|_2 \geq \frac{1}{\varepsilon} \right\}.$$

a. Prove that

$$\sigma_\varepsilon(A) = \bigcup_{\|B\|_2 \leq \varepsilon} \sigma(A + B).$$

b. Prove also that

$$\sigma_\varepsilon(A) \subset \{z \in \mathbb{C}; \text{dist}(z; \mathcal{H}(A)) \leq \varepsilon\},$$

where  $\mathcal{H}(A)$  is the numerical range of  $A$ .

**Note:** the notion of pseudo-spectrum is fundamental in several scientific domains, including dynamical systems, numerical analysis, and quantum mechanics (in semiclassical analysis, one speaks of *quasi-modes*). The reader interested in this subject should consult the book by Trefethen and Embree [38].

29. We recall that the numerical radius of  $A \in \mathbf{M}_n(\mathbb{C})$  is defined by

$$w(A) := \sup\{|z|; z \in \mathcal{H}(A)\} = \sup\{|x^*Ax|; x \in \mathbb{C}^n, \|x\|_2 = 1\}.$$

Prove that

$$w(A) \leq \|A\|_2 \leq 2w(A).$$

**Hint:** Use the polarization principle to prove the second inequality.

30. Let  $A \in \mathbf{M}_n(\mathbb{C})$  be a nilpotent matrix of order two:  $A^2 = 0_n$ .

a. Using standard properties of the norm  $\|\cdot\|_2$ , verify that  $\|M\|_2^2 \leq \|MM^* + M^*M\|_2$  for every  $M \in \mathbf{M}_n(\mathbb{C})$ .

b. When  $k$  is a positive integer, compute  $(AA^* + A^*A)^k$  in closed form. Deduce that

$$\|AA^* + A^*A\|_2 \leq 2^{1/k} \|A\|_2^2.$$

c. Passing to the limit as  $k \rightarrow +\infty$ , prove that

$$\|A\|_2 = \|AA^* + A^*A\|_2^{1/2}. \quad (7.7)$$