

# Appendix F

## Answers to some exercises

### F.1 Chapter 1

1.4 Existence: since  $f$  is continuous we are guaranteed that  $\|f\|_1$  exists

*positive homogeneous:* for every scalar  $\alpha$  and function  $f$  we have  $\|\alpha f\|_1 = \int_a^b |\alpha f(t)| dt = |\alpha| \int_a^b |f(t)| dt = |\alpha| \|f\|_1$ .

*triangle inequality:* for every two functions  $f, g$  we have  $\|f + g\|_1 = \int_a^b |f(t) + g(t)| dt \leq \int_a^b |f(t)| + |g(t)| dt = \|f\|_1 + \|g\|_1$ .

*positive definite:* if  $f$  is not the zero function, then,  $f(t_0) \neq 0$  for some  $t_0 \in [a, b]$ . Then because of continuity  $|f(t)| > |f(t_0)|/2$  for all  $t$  in some small enough neighborhood of  $t_0$ . Then  $\|f\|_1 > 0$ .

1.5 Yes because:

*Existence:*  $\|x\|_a + \|x\|_b$  is well defined because both  $\|x\|_a$  and  $\|x\|_b$  are defined (because they are norms).

*Positive homogeneous:*  $\|\alpha x\| := \|\alpha x\|_a + \|\alpha x\|_b = |\alpha| \|x\|_a + |\alpha| \|x\|_b = |\alpha| (\|x\|_a + \|x\|_b) = |\alpha| \|x\|$ . So it is positive homogeneous.

*Triangle inequality:*  $\|x + y\| := \|x + y\|_a + \|x + y\|_b \leq (\|x\|_a + \|y\|_a) + (\|x\|_b + \|y\|_b) = (\|x\|_a + \|x\|_b) + (\|y\|_a + \|y\|_b) = \|x\| + \|y\|$ . So triangle inequality holds.

*Positive definite:* if  $x \neq 0$  then  $\|x\| := \|x\|_a + \|x\|_b > 0$  because  $\|x\|_a > 0$  (since  $\|\cdot\|_a$  is a norm.) So all axioms of norm hold.

1.8 (By the way: this assumes that the cartesian product is equipped with the natural vector addition and scalar product:  $(x_a, x_b) + (y_a, y_b) = (x_a + y_a, x_b + y_b)$  and  $\alpha(x_a, x_b) = (\alpha x_a, \alpha x_b)$ .)

*Existence:* since  $\|\cdot\|_A$  and  $\|\cdot\|_B$  are norms, they are well defined for all its elements. Then  $\max(\|x_a\|_A, \|x_b\|_B)$  is also well defined (finite).

*Positive homogeneous:*  $\|\alpha(x_a, x_b)\| = \|(\alpha x_a, \alpha x_b)\| = \max(\|\alpha x_a\|_A, \|\alpha x_b\|_B)$  which by the norm properties of  $\|\cdot\|_A, \|\cdot\|_B$  equals  $|\alpha| \max(\|x_a\|_A, \|x_b\|_B)$ . So it is positive homogeneous.

*Triangle inequality:*  $\|(x_a, x_b) + (y_a, y_b)\| = \|(x_a + y_a, x_b + y_b)\| = \max(\|x_a + y_a\|_A, \|x_b + y_b\|_B)$ . Case 1: suppose that  $\|x_a + y_a\|_A \geq \|x_b + y_b\|_B$ . Then the above

gives  $\|(x_a, x_b) + (y_a, y_b)\| = \|x_a + y_a\|_A \leq \|x_a\|_A + \|y_a\|_A \leq \max(\|x_a\|_A, \|x_b\|_B) + \max(\|x_b\|_B, \|y_b\|_B) = \|(x_a, x_b)\| + \|(y_a, y_b)\|$ . This is the triangle inequality. The other case (so  $\|x_a + y_a\|_A \leq \|x_b + y_b\|_B$ ) results in the same triangle inequality.

*Positive definiteness:* if  $(x_a, x_b)$  is nonzero then at least one of the two is nonzero, so then  $\|(x_a, x_b)\| = \max(\|x_a\|_A, \|x_b\|_B) > 0$ .

1.9 yes...

1.10 On  $\mathbb{R}$  we take the absolute value as norm (Exercise 1.6 explains that this is, essentially, without loss of generality). Suppose  $f_n$  is Cauchy. Then  $\forall \epsilon > 0$  there is an  $N > 0$  such that  $\|f_n - f_m\| < \epsilon$  for all  $n > N$ . Then  $a_n := \|f_n\|$  by the reverse triangle inequality satisfies  $|a_n - a_m| = |\|f_n\| - \|f_m\|| < \|f_n - f_m\| < \epsilon$  for all  $n, m > N$ . Hence  $a_n$  is a Cauchy sequence (in  $\mathbb{R}$  with absolute value as norm).

1.11 Take an  $N$ . For all  $n, m > N$  we have that  $f_n(t) - f_m(t) = 0$  for all  $t > 1/N$ . Hence for all  $n, m > N$  we have

$$\|f_n - f_m\|_2^2 \leq \int_0^{1/N} |f_n(t) - f_m(t)|^2 dt$$

All functions  $f_n(t), f_m(t)$  are nonnegative and bounded from above by  $\frac{1}{t^{1/5}} = t^{-0.2}$  so we have for all  $n, m > N$  that

$$\|f_n - f_m\|_2^2 \leq \int_0^{1/N} (2t^{-0.2})^2 dt = \left[ 4 \frac{t^{0.6}}{0.6} \right]_0^{1/N} = \frac{4}{0.6} \frac{1}{N^{0.6}}$$

This converges to zero as  $N \rightarrow \infty$  so  $f_n$  is a Cauchy sequence.

1.12 Suppose  $f_n$  is Cauchy. Take an  $\epsilon > 0$  and let  $N$  be such that  $\|f_n - f_m\| < \epsilon$  for all  $n > N$ . This implies that  $\|f_n\| \leq \|f_m\| + \epsilon$  for all  $n, m > N$ . For fixed such  $m$  and  $\epsilon$  this shows that  $\|f_n\|$  for all  $n > N$  is bounded by  $M := \|f_m\| + \epsilon$ . Then  $f_n$  for all  $n$  is bounded by

$$\|f_n\| \leq \max(\|f_1\|, \|f_2\|, \dots, \|f_N\|, M) < \infty$$

The upper bound does not depend on  $n$ , implying that  $f_n$  is bounded.

1.13  $\|A(x)\|_2 = \left\| \begin{bmatrix} 2x_1 \\ 3x_2 \end{bmatrix} \right\|_2 = \sqrt{4|x_1|^2 + 9|x_2|^2} \leq \sqrt{9|x_1|^2 + 9|x_2|^2} = 3\|x\|_2$ . So  $\|A\| \leq 3$ . It is in fact equal to 3 because for  $x = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  have have  $\|x\|_2 = 1$  and  $\|A(x)\|_2 = \left\| \begin{bmatrix} 0 \\ 3 \end{bmatrix} \right\|_2 = 3$  (so  $\|A(x)\|_2 / \|x\|_2 = 3$  for this  $x$ )

1.14 If  $v_i$  is an eigenvector with eigenvalue  $\lambda_i$  then

$$\frac{\|A(v_i)\|}{\|v_i\|} = \frac{\|\lambda_i v_i\|}{\|v_i\|} = |\lambda_i|.$$

Since the eigenvectors are only a subset of  $\mathbb{V}$  we have

$$\|A\| := \sup_{v \in \mathbb{V}} \frac{\|A(v)\|}{\|v\|} \geq \sup_{v=v_i} \frac{\|A(v)\|}{\|v\|} = \max_i |\lambda_i|.$$

- 1.15 (a)  $\|K(u)\| = \|(0, u_1, u_2, \dots)\|_2 = \sqrt{0^2 + u_1^2 + u_2^2 + \dots} = \|u\|$  so  $\|K\| = 1$
- (b)  $\|L(u)\| = \|(u_2, u_3, \dots)\|_2 = \sqrt{u_2^2 + u_3^2 + \dots} \leq \sqrt{u_1^2 + u_2^2 + u_3^2 + \dots} = \|u\|$  so  $\|L\| \leq 1$ .  
 Now let  $u = (0, 1, 0, 0, \dots)$ . Then  $L(u) = (1, 0, 0, \dots)$  so  $\|L(u)\|_2 = \|u\|_2 = 1$  in this case, showing that  $\|L\| \geq 1$ . So we have  $\|L\| = 1$
- (c)  $\|M(u)\|_2 = \sqrt{\sum_{k=1}^{\infty} (2 - 1/k)^2 u_k^2} \leq \sqrt{\sum_{k=1}^{\infty} 2^2 u_k^2} = 2\|u\|$ . So  $\|M\| \leq 2$ .  
 Now consider the “sequence of sequences”  $e_n := (0, \dots, 0, 0, 1, 0, 0, \dots)$  (meaning  $e_n$  has a 1 at its  $n$ th position and all other entries are zero). Then  $M(e_n) = (2 - 1/n)e_n$  implying that  $\|M\| \geq \|M(e_n)\|/\|e_n\| = (2 - 1/n)$ . In the limit as  $n \rightarrow \infty$  this lower bound for  $\|M\|$  converges to 2. So  $\|M\| = 2$ .

1.16 -

- 1.17 Exploit the triangle inequality of norm:  $\|cA(f)\|_2 = \sqrt{\int_{-\infty}^{\infty} |f(t) + f(-t)|^2 dt} = \|f(t) + f(-t)\|_2 \leq \|f(t)\|_2 + \|f(-t)\|_2 = 2\|f\|_2$ . So  $\|A\| \leq 2$ . For, say,

$$f(t) = \begin{cases} 1 & \text{if } t \in [0, 5] \\ 0 & \text{elsewhere} \end{cases}$$

we have  $\|f\|_2 = 5$  and  $\|A(f)\|_2 = \sqrt{\int_{-5}^5 1^2 dt} = 10$ . So  $\|A(f)\|/\|f\|_2 = 2$  here. Combined with  $\|A\| \leq 2$  shows that  $\|A\| = 2$ .

1.18

1.20

- 1.22 (a) Suppose  $f$  and  $g$  are two limits of a convergent  $f_n$ . Since convergent, for every  $\epsilon > 0$  there are  $N_f, N_g$  such that  $\|f - f_n\| < \epsilon \forall n > N_f$  and  $\|g - f_n\| < \epsilon \forall n > N_g$ . Then  $\|f - g\| = \|(f - f_n) - (g - f_n)\| \leq \|f - f_n\| + \|g - f_n\| < 2\epsilon$  for all  $n > \max(N_f, N_g)$ . Since  $\epsilon$  is arbitrary we must have  $\|f - g\| = 0$ . By the third axiom norm this means  $f = g$ .
- (b) Let  $f = \lim_{n \rightarrow \infty} f_n$  as an element of  $\mathbb{X}$ . If  $f \in \mathbb{Y}$  then clearly  $f_n$  converges in  $\mathbb{Y}$ . If  $f \notin \mathbb{Y}$  then  $f_n$  cannot converge in  $\mathbb{Y}$  because if it would then this limit (call it  $y$ ) would mean that  $f_n$  has two different limits ( $f$  and  $y$ ) in  $\mathbb{X}$ . Not possible.
- (c) Since  $\mathcal{L}^1([-1, 1]; \mathbb{R})$  (with its 1-norm) is Banach, and  $f_n$  is a Cauchy sequence in the 1-norm, it follows that  $f := \lim_{n \rightarrow \infty} f_n$  is  $\mathcal{L}^1([-1, 1]; \mathbb{R})$  and is unique. Clearly this is  $f(t) = 0$  for  $t < 0$  and  $f(t) = 1$  for all  $t > 1$ . This function is (essentially) not continuous. Apply the previous part (using that  $\mathcal{C}([-1, 1]; \mathbb{R}) \subset \mathcal{L}^1([-1, 1]; \mathbb{R})$ ).

1.23 -