

# Functional- and Fourieranalysis

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This scriptum is by no means intended to be original: It is a note on the things I do in the lecture. In some parts, it is very similar to the excellent books by Christian Blatter [?] and Hans Wilhelm Alt [?]. In particular, I try to adapt – if possible – to Blatter’s notation, making it easier for readers to read the book. (This book can be found on Blatter’s website <https://people.math.ethz.ch/~blatter/dlp.html> in German, it is also available in UHasselt’s library in English.)

This scriptum is only intended for use in my 2020 class!

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## 1 Preliminaries, Banach- and Hilbert spaces

In this lecture, we consider Banach- and Hilbert spaces. Those spaces are always vector spaces over a field  $\mathbb{K} \in \{\mathbb{R}, \mathbb{C}\}$  with a norm and some special properties.

**Definition 1** (Normed space). *Let  $\mathcal{X}$  be a vector space. A function  $\|\cdot\| : \mathcal{X} \rightarrow \mathbb{R}^{\geq 0}$  is called a norm, if there holds for all  $x, y \in \mathcal{X}$ ,  $\lambda \in \mathbb{K}$ ,*

- $\|\lambda x\| = |\lambda| \|x\|$  (homogeneity),
- $\|x\| = 0$  if and only if  $x = 0$ ,
- $\|x + y\| \leq \|x\| + \|y\|$  (triangle inequality).

*The pairing  $(\mathcal{X}, \|\cdot\|)$  is called a normed space.*

**Example 1.** *The following pairs are normed spaces:*

- $(\mathbb{R}^n, \|\cdot\|_p)$  for  $p \in [1, \infty]$  with the usual definition  $\|x\|_p := \sqrt[p]{\sum_{i=1}^n |x_i|^p}$  for  $p < \infty$  and  $\|x\|_\infty := \max_{1 \leq i \leq n} |x_i|$ . It is easy to show that these norms fulfill the homogeneity property and  $\|x\|_p = 0$  if and only if  $x = 0$ , however, save for  $p \in \{1, \infty\}$ , the triangle inequality is difficult to show, it will be shown below, see Lemma 1.
- The space  $C^0([a, b])$  of continuous functions  $[a, b] \rightarrow \mathbb{R}$ , equipped with the supremum-norm

$$\|f\|_\infty := \sup_{x \in [a, b]} |f(x)|$$

*is a normed space. Addition and scalar multiplication are to be understood componentwise, so  $f + g$  is the function  $x \mapsto f(x) + g(x)$ .*

- The space of functions that are Lebesgue-integrable to power  $p$ ,

$$L^p([a, b]) := \{f : [a, b] \rightarrow \mathbb{R} \mid f \text{ is integrable, } \int_a^b |f(x)|^p dx < \infty\},$$

together with the norm

$$\|f\|_p := \sqrt[p]{\int_a^b |f(x)|^p dx}$$

is a normed space if  $1 \leq p \leq \infty$ . (Note that for  $p \equiv \infty$ , the norm has to be replaced by the sup-norm.) Note that the notation  $\|\cdot\|_p$  is ambiguous, depending on whether  $\mathbb{R}^n$  or  $L^p$  is considered.

This course on *functional analysis* is the extension of linear algebra (which dealt with finite-dimensional spaces) to infinite-dimensional spaces. Some unexpected behavior shows up then. Before we show some examples, we start with the following important theorems:

**Theorem 1** (Young's inequality). *Let  $p, q \in \mathbb{R}^{>1}$  such that  $\frac{1}{p} + \frac{1}{q} = 1$ , and let  $a, b \in \mathbb{R}^{\geq 0}$ . Then, there holds*

$$ab \leq \frac{a^p}{p} + \frac{b^q}{q}.$$

*Proof.* We may assume that  $a, b > 0$ , as the inequality becomes trivial otherwise. The logarithm function is concave ( $\ln''(x) = -\frac{1}{x^2} < 0$ ), so there holds

$$\ln(x + \tau(y - x)) \geq (1 - \tau) \ln(x) + \tau \ln(y), \quad \tau \in [0, 1].$$

From this and some elementary rules on the logarithm, we get (set  $\tau := \frac{1}{q}$  and hence  $(1 - \tau) = \frac{1}{p}$ )

$$\ln(ab) = \ln(a) + \ln(b) = \frac{1}{p} \ln(a^p) + \frac{1}{q} \ln(b^q) \leq \ln\left(\frac{a^p}{p} + \frac{b^q}{q}\right).$$

Monotonicity of  $\ln$  concludes the proof. □

From this, we can derive the discrete Hölder inequality:

**Theorem 2** (Hölder's inequality). *Let  $p, q \in \mathbb{R}^{\geq 1} \cup \{\infty\}$  such that  $\frac{1}{p} + \frac{1}{q} = 1$ . (We set  $\frac{1}{\infty} := 0$  for this.) Then, there holds for all  $x, y \in \mathbb{R}^n$  that*

$$\sum_{i=1}^n |x_i| |y_i| \leq \|x\|_p \|y\|_q.$$

*Proof.* We only consider  $x, y \neq 0$ . The claim is easy to prove for  $p = 1$  and  $q = \infty$  (or the other way around), so we assume that  $p, q \notin \{1, \infty\}$ . Then, one can write

$$\sum_{i=1}^n \frac{|x_i|}{\|x\|_p} \frac{|y_i|}{\|y\|_q} \stackrel{\text{Young}}{\leq} \sum_{i=1}^n \left( \frac{|x_i|^p}{p \|x\|_p^p} + \frac{|y_i|^q}{q \|y\|_q^q} \right) = \frac{1}{p} + \frac{1}{q} = 1.$$

Multiplying by  $\|x\|_p \|y\|_q$  concludes the proof. □

Let us, for the sake of completeness, give the Hölder inequality also for Lebesgue functions:

**Theorem 3** (Hölder's inequality for Lebesgue functions). *Let  $p, q \in \mathbb{R}^{\geq 1} \cup \{\infty\}$  such that  $\frac{1}{p} + \frac{1}{q} = 1$ . Note the above convention for  $p$  or  $q$  equal to infinity. Let  $u \in L^p(\Omega)$  and  $v \in L^q(\Omega)$ , where  $\Omega \subset \mathbb{R}^n$  is an open set. Then, there holds*

$$\int_{\Omega} |u| |v| dx \leq \|u\|_p \|v\|_q.$$

An important application of Hölder's inequality is the triangle inequality:

**Lemma 1** (Triangle inequality of  $\|\cdot\|_p$ ). *For  $x, y \in \mathbb{R}^n$ , there holds*

$$\|x + y\|_p \leq \|x\|_p + \|y\|_p$$

for  $p \in [1, \infty]$ .

*Proof.* The proof is easy for  $p \in \{1, \infty\}$ , as one only has to rely on the triangle inequality for the absolute value. Hence, we restrict ourselves to  $1 < p < \infty$ . Note that we will define  $q := (1 - \frac{1}{p})^{-1} = \frac{p}{p-1}$ , so that  $\frac{1}{p} + \frac{1}{q} = 1$ . There holds:

$$\begin{aligned} \|x + y\|_p^p &= \sum_{i=1}^n |x_i + y_i|^p = \sum_{i=1}^n |x_i + y_i| |x_i + y_i|^{p-1} \\ &\leq \sum_{i=1}^n |x_i| |x_i + y_i|^{p-1} + \sum_{i=1}^n |y_i| |x_i + y_i|^{p-1} \\ &\stackrel{\text{Hölder}}{\leq} \|x\|_p \left( \sum_{i=1}^n |x_i + y_i|^{(p-1) \cdot q} \right)^{1/q} + \|y\|_p \left( \sum_{i=1}^n |x_i + y_i|^{(p-1) \cdot q} \right)^{1/q} \\ &= (\|x\|_p + \|y\|_p) \left( \sum_{i=1}^n |x_i + y_i|^{(p-1) \cdot q} \right)^{1/q} \\ &= (\|x\|_p + \|y\|_p) \left( \sum_{i=1}^n |x_i + y_i|^p \right)^{\frac{p-1}{p}} = (\|x\|_p + \|y\|_p) \|x + y\|_p^{p-1} \end{aligned}$$

If  $x + y \neq 0$ , then division by  $\|x + y\|_p^{p-1}$  gives the desired result. For  $x + y = 0$ , the inequality is trivial.  $\square$

Having discussed these very fundamental properties, we will now look into infinite-dimensional spaces.

**Example 2.** *We give some examples of infinite-dimensional spaces.*

- By  $\mathbb{K}^{\mathbb{N}}$ , we denote the set of all sequences with values in  $\mathbb{K}$ , so

$$\mathbb{K}^{\mathbb{N}} := \{(a_i)_{i \in \mathbb{N}} \mid a_i \in \mathbb{K} \forall i \in \mathbb{N}\}.$$

Similar to norms on  $\mathbb{R}^n$ , define for  $p \in [1, \infty)$  the functions

$$\|a\|_{l^p} := \sqrt[p]{\sum_{i \in \mathbb{N}} |a_i|^p}, \quad \|a\|_{l^\infty} := \sup_{i \in \mathbb{N}} |a_i|.$$

Note that those are not norms on  $\mathbb{K}^{\mathbb{N}}$ . (Why not?) They are norms on the subspaces

$$l^p(\mathbb{K}) := \{a \in \mathbb{K}^{\mathbb{N}} \mid \|a\|_{l^p} < \infty\},$$

$l^p(\mathbb{K})$  is thus a normed space. For  $i \in \mathbb{N}$ , define  $e_i \in \mathbb{K}^{\mathbb{N}}$  as the sequence having one at position  $i$ , and zeros elsewhere. Obviously,  $e_i \in l^p(\mathbb{K})$  for any  $p \in [1, \infty]$ , and all  $e_i$  are linearly independent. Therefore,  $l^p(\mathbb{K})$  is indeed an infinite-dimensional space.

- The space  $C^0([a, b])$  is also infinite-dimensional. To see this, choose  $n \in \mathbb{N}$  and select  $n$  pairwise distinct points  $x_i \in [a, b]$ ,  $1 \leq i \leq n$ . Denote  $\varepsilon := \frac{1}{2} \min_{i \neq j} |x_i - x_j| > 0$ . There exist continuous functions  $f_i$  with the following properties:

- $\text{supp}(f_i) \subset (x_i - \varepsilon, x_i + \varepsilon)$ ,
- $f_i(x_i) = 1$ .

(Show this!) As a consequence,  $f_i(x_j) = 0$  for any  $j \neq i$ . Also these  $f_i$  must be linearly independent. As  $n \in \mathbb{N}$  was arbitrary,  $C^0([a, b])$  cannot have a finite dimension.

**Example 3.** Let  $\mathcal{X}$  be a finite-dimensional vector space,  $\dim(\mathcal{X}) = n$ . We know that there exists a basis  $e_i \in \mathcal{X}$ ,  $i = 1, \dots, n$ , such that any  $x \in \mathcal{X}$  can be written as a linear combination of these  $e_i$ . One consequence is that any two norms  $\|\cdot\|$  and  $\|\cdot\|_*$  that can be defined on  $\mathcal{X}$  are equivalent. As a reminder, being equivalent means that there are constants  $c, C \in \mathbb{R}^{>0}$  such that

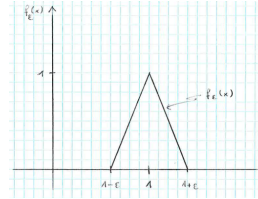
$$c\|x\| \leq \|x\|_* \leq C\|x\|, \quad \forall x \in \mathcal{X}.$$

Norms are not necessarily equivalent any more in infinite-dimensional spaces! Consider  $C^0([a, b])$  with the supremum-norm  $\|\cdot\|_\infty$  and the integral norm  $\|\cdot\|_1$ ,  $\|f\|_1 := \int_a^b |f|dx$ . There holds

$$\|f\|_1 \leq |b - a| \|f\|_\infty,$$

however,  $c\|f\|_\infty \leq \|f\|_1$  does not hold for any  $c \in \mathbb{R}^{>0}$ . To see this, consider for simplicity  $[a, b] = [0, 2]$  (idea is the same on any other  $[a, b]$ ), and define the continuous, piecewise linear parametric function

$$f_\varepsilon(x) := \begin{cases} \frac{x-1+\varepsilon}{\varepsilon}, & x \in [1-\varepsilon, 1], \\ \frac{1-x+\varepsilon}{\varepsilon}, & x \in (1, 1+\varepsilon], \\ 0, & \text{otherwise.} \end{cases}$$



For any  $0 < \varepsilon < 1$ , there holds  $\|f_\varepsilon\|_\infty = 1$ , but  $\|f_\varepsilon\|_1 = \varepsilon$ .

As a reminder: A sequence  $(a_i)_{i \in \mathbb{N}}$  is called a Cauchy sequence if for any  $\varepsilon > 0$ , there exists  $n_0 \in \mathbb{N}$ , such that for all  $i, j \geq n_0$ , there holds

$$|a_i - a_j| \leq \varepsilon.$$

Cauchy series do not necessarily have to converge. As an example, consider the space  $\mathbb{Q}$  and the sequence  $a_i$  with  $i \mapsto \left(1 + \frac{1}{i}\right)^i \in \mathbb{Q}$ . We know that (in  $\mathbb{R}$ ) this sequence converges to  $e$ , so there exists for any  $\varepsilon > 0$  an  $n_0$  such that  $|e - a_i| \leq \frac{\varepsilon}{2}$  for any  $i \geq n_0$ . Consequently,

$$|a_i - a_j| \leq |a_i - e| + |a_j - e| \leq \varepsilon, \quad i, j \geq n_0$$

and  $(a_i)$  is a Cauchy sequence in  $\mathbb{Q}$ . Still, it has no limit in  $\mathbb{Q}$ , because  $\mathbb{Q}$  is not complete:

**Definition 2** (Completeness). A normed space  $(\mathcal{X}, \|\cdot\|)$  is called complete, if every Cauchy sequence has a limit in  $\mathcal{X}$ .

Having said this, we come to the very important definition of what is a Banach space.

**Definition 3** (Banach space). A normed vector space  $(\mathcal{X}, \|\cdot\|)$  that is complete (w.r.t. to the norm) is called a Banach space.

**Remark 1.** Let us note that many of the completeness proofs have the following structure:

- First, one identifies a suitable 'weaker' limit. (These could be pointwise limits for functions, or element-wise limits for sequences. )

- Then, one shows that this 'weaker' limit is an actual limit, and also that it belongs to  $\mathcal{X}$ .

**Example 4.** Here, we show some examples of (non-)complete vector spaces.

- We begin with the space  $l^p(\mathbb{K})$  and the associated norm  $\|\cdot\|_p$ . (We consider  $p < \infty$ ,  $p = \infty$  is similar!) This is a complete normed space. To see this, assume that we have a sequence  $(x_n)_{n \in \mathbb{N}} \subset l^p(\mathbb{K})$ . (So the  $x_n \in l^p(\mathbb{K})$  are themselves sequences!) Being a Cauchy sequence means that

$$\sum_{i \in \mathbb{N}} |x_{n,i} - x_{m,i}|^p = \|x_n - x_m\|_p^p \leq \varepsilon,$$

for all  $n, m$  above some given threshold. (Note that power of  $p$ . We will do that frequently in order not having to work with the root.) In particular, this implies that  $(x_{n,i})_{n \in \mathbb{N}} \subset \mathbb{K}$  is a Cauchy sequence. Even more, for all  $i$  and a given  $\varepsilon$ , we can find  $n_0$  such that for all  $n, m > n_0$ , there holds

$$|x_{n,i} - x_{m,i}| \leq \varepsilon_i$$

such that  $\sum_{i \in \mathbb{N}} \varepsilon_i^p \leq \varepsilon$ . Because  $\varepsilon_i^p \leq \varepsilon$  and  $n_0$  does not depend on  $i$ , this is of course uniform convergence. As a field is always complete, there exists a limit  $\bar{x}_i$  to the sequence  $(x_{n,i})_{n \in \mathbb{N}}$ . Passing to the limit in the above inequality, one can also conclude that

$$|x_{n,i} - \bar{x}_i| \leq \varepsilon_i.$$

Define the sequence  $(\bar{x}_i)_{i \in \mathbb{N}}$ . We have to show that it is in  $l^p(\mathbb{K})$  and that it is the limit of  $(x_n)_{n \in \mathbb{N}}$ . So, there holds due to La. 1 (assume  $n$  is chosen sufficiently large)

$$\left( \sum_{i \in \mathbb{N}} |\bar{x}_i|^p \right)^{\frac{1}{p}} = \left( \sum_{i \in \mathbb{N}} |x_{n,i} - x_{n,i} + \bar{x}_i|^p \right)^{\frac{1}{p}} \leq \left( \sum_{i \in \mathbb{N}} |x_{n,i} - \bar{x}_i|^p \right)^{\frac{1}{p}} + \left( \sum_{i \in \mathbb{N}} |x_{n,i}|^p \right)^{\frac{1}{p}} < \infty,$$

we therefore have  $(\bar{x}_i)_{i \in \mathbb{N}} \in l^p(\mathbb{K})$ . Furthermore,

$$\sum_{i \in \mathbb{N}} |x_{n,i} - \bar{x}_i|^p \leq \sum_{i \in \mathbb{N}} \varepsilon_i^p \leq \varepsilon,$$

so  $(\bar{x}_i)_{i \in \mathbb{N}}$  is in fact the limit of  $(x_n)_{n \in \mathbb{N}}$ .

- We now look at  $C^0([0, 1])$  paired with two different norms.
  - First, we take the  $\|\cdot\|_\infty$  norm. Then, the space is complete. To see this, consider a sequence of continuous functions,  $(f_n)_{n \in \mathbb{N}} \subset C^0([0, 1])$ . Assuming that it is Cauchy means that

$$|f_n(x) - f_m(x)| \leq \|f_n - f_m\|_\infty \leq \varepsilon$$

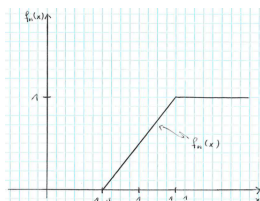
for  $n, m$  sufficiently large and  $x \in [0, 1]$ . This implies that for a fixed  $x$ ,  $(f_n(x))_{n \in \mathbb{N}}$  converges to some limit  $f(x)$ . We have to show that this function is then again in  $C^0([0, 1])$ , we only have to show continuity of course. We compute, for  $|\delta|$  sufficiently small such that  $x + \delta \in [0, 1]$ , and  $x \in (0, 1)$ ,

$$\begin{aligned} |f(x + \delta) - f(x)| &= |f(x + \delta) - f_n(x + \delta) + f_n(x + \delta) - f_n(x) + f_n(x) - f(x)| \\ &\leq |f(x + \delta) - f_n(x + \delta)| + |f_n(x + \delta) - f_n(x)| + |f_n(x) - f(x)| \\ &\leq \varepsilon + |f_n(x + \delta) - f_n(x)| + \varepsilon. \end{aligned}$$

We have assumed that  $n$  is so large that  $|f(x) - f_n(x)| \leq \varepsilon$ . As  $f_n$  is continuous,  $|f_n(x + \delta) - f_n(x)| \rightarrow 0$  with  $\delta \rightarrow 0$ .  $f(x)$  is therefore in  $C^0([0, 1])$  and the space is complete.

- Taking the  $\|\cdot\|_1$ -norm shows a different picture, the space  $C^0([0,1])$  will not be complete anymore under this norm. Take the sequence of functions

$$f_n(x) := \begin{cases} 0, & x < \frac{1}{2} - \frac{1}{2n} \\ n(x - \frac{1}{2}) + \frac{1}{2}, & \frac{1}{2} - \frac{1}{2n} \leq x \leq \frac{1}{2} + \frac{1}{2n} \\ 1, & \text{otherwise} \end{cases}$$



The  $f_n$  are continuous and form a Cauchy sequence w.r.t. to the  $\|\cdot\|_1$  norm (try this as an exercise!). The limit, however, is the function

$$f(x) := \begin{cases} 0, & x < \frac{1}{2} \\ 1, & x > \frac{1}{2} \end{cases}$$

which is not continuous any more.

Most of the examples we will be considering in this lecture are indeed Banach spaces. Even more, they are typically Banach spaces with a norm induced by a scalar product.

**Definition 4** (Scalar product). A function  $(\cdot, \cdot) : \mathcal{X} \times \mathcal{X} \rightarrow \mathbb{K}$  is called a scalar product if it is sesquilinear, which means for  $\alpha, \beta \in \mathbb{K}$  and  $x, y, z \in \mathcal{X}$  that

$$\begin{aligned} (\alpha x + \beta y, z) &= \alpha(x, z) + \beta(y, z), \\ (z, \alpha x + \beta y) &= \overline{\alpha}(z, x) + \overline{\beta}(z, y), \end{aligned}$$

and if additionally, there holds:

- $(u, v) = \overline{(v, u)}$ . (This is the complex conjugate; in case  $\mathbb{K} = \mathbb{R}$  there holds  $(u, v) = (v, u)$  of course.)
- $(u, u) > 0$  for all  $u \in \mathcal{X} \setminus \{0\}$ .

**Remark 2.** • Note that  $(0, 0) = 0$  follows easily due to linearity.

- The condition  $(u, u) > 0$  makes only sense if  $(u, u) \in \mathbb{R}$ . This, however, is a consequence of  $(u, v) = \overline{(v, u)}$  and hence  $(u, u) = \overline{(u, u)} \in \mathbb{R}$ .

**Lemma 2.** If the vector space  $\mathcal{X}$  is equipped with a scalar product  $(\cdot, \cdot)$ , then this scalar product induces a norm through

$$\|\cdot\| : x \mapsto \sqrt{(x, x)}.$$

Before proving this, we will first discuss Cauchy-Schwarz's inequality. For vectors  $u, v \in \mathbb{R}^2$ , it is known that

$$(u, v) = \|u\| \|v\| \cos(\theta),$$

with  $\theta$  the angle between  $u$  and  $v$ . Consequently, as  $|\cos(\cdot)| \leq 1$ , there holds  $|(u, v)| \leq \|u\| \|v\|$ . The extension of this to general linear spaces is the core of the following theorem:

**Theorem 4** (Cauchy-Schwarz inequality). *Let the vector space  $\mathcal{X}$  be equipped with a scalar product  $(\cdot, \cdot)$ , and let  $\|\cdot\|$  be the induced norm. Then, there holds*

$$|(u, v)| \leq \|u\| \|v\|, \quad \forall u, v \in \mathcal{X}.$$

*There is equality if and only if  $u$  and  $v$  are linearly dependent.*

*Proof.* The theorem is obviously true if  $u$  or  $v$  are zero. Let us therefore assume  $u \neq 0 \neq v$ . We define the orthogonal projection of  $v$  onto  $u$  as  $\frac{(v, u)}{\|u\|^2} u$ . Then, there holds

$$\begin{aligned} 0 \leq \left\| v - \frac{(v, u)}{\|u\|^2} u \right\|^2 &= (v, v) + \frac{|(v, u)|^2 (u, u)}{\|u\|^4} - \left( v, \frac{(v, u)}{\|u\|^2} u \right) - \left( \frac{(v, u)}{\|u\|^2} u, v \right) \\ &= \|v\|^2 + \frac{|(v, u)|^2}{\|u\|^2} - 2 \frac{|(v, u)|^2}{\|u\|^2} = \|v\|^2 - \frac{|(v, u)|^2}{\|u\|^2}. \end{aligned}$$

(You should try and do the intermediate steps. Note that  $(\cdot, \cdot)$  is sesquilinear.) Multiplied by  $\|u\|^2$ , this yields

$$|(v, u)|^2 \leq \|v\|^2 \|u\|^2,$$

which is the inequality claimed. There is equality if and only if  $v - \frac{(v, u)}{\|u\|^2} u = 0$ , which of course means that  $u$  and  $v$  are linearly dependent.  $\square$

**Corollary 1.** *Assume the same conditions as in Thm. 4. Then, there exists a  $\theta$  (unique in  $[0, \pi]$ ) such that*

$$(u, v) = \|u\| \|v\| \cos(\theta).$$

We now prove that  $x \mapsto \sqrt{(x, x)}$  does indeed define a norm.

*Proof of La. 2.* The only non-trivial part to show is the triangle inequality, so let  $x, y \in \mathcal{X}$ , then

$$\begin{aligned} \|x + y\|^2 &= (x + y, x + y) = \|x\|^2 + (x, y) + (y, x) + \|y\|^2 \\ &= \|x\|^2 + \|y\|^2 + 2 \operatorname{Re}((x, y)) \\ &\leq \|x\|^2 + \|y\|^2 + 2|(x, y)| \\ &\leq \|x\|^2 + \|y\|^2 + 2\|x\| \|y\| = (\|x\| + \|y\|)^2. \end{aligned}$$

This concludes the proof.  $\square$

Having said this, we can define Hilbert spaces:

**Definition 5** (Hilbert space). *A Banach space with scalar-product-induced norm is called a Hilbert space.*

**Example 5.** *Let us show two examples of Hilbert spaces.*

- Obviously,  $\mathbb{R}^n$  together with the 'standard' scalar product  $(x, y) := \sum_{i=1}^n x_i y_i$  is a Hilbert space. The induced norm is  $\|\cdot\|_2$ . Note that for any  $p \neq 2$ ,  $\|\cdot\|_p$  is not a norm induced by a scalar product. This will become clear later.
- Because the following space will be of utmost importance to Fourier analysis, we will have a definition within an example:

**Definition 6.** We define  $L^2_{\circ}$  to be the space of square-integrable periodic functions with periodicity  $2\pi$ , i.e., for any  $f \in L^2_{\circ}$ , there holds  $f(x + 2\pi) = f(x)$  almost everywhere, and the integral

$$\|f\|_{L^2_{\circ}}^2 := \frac{1}{2\pi} \int_{-\pi}^{\pi} |f(x)|^2 dx$$

is finite. (We use the scaling of  $\frac{1}{2\pi}$  for our convenience, in Fourier analysis, it will turn out that then, a lot of constants are one.)

Then,  $L^2_{\circ}$  is a Hilbert space together with the scalar product

$$(f, g)_{L^2_{\circ}} := \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) \bar{g}(x) dx.$$

**Lemma 3** (Parallelogram identity). *Let  $\mathcal{X}$  be a Hilbert space. Then, there holds for all  $u, v \in \mathcal{X}$ ,*

$$\|u + v\|^2 + \|u - v\|^2 = 2\|u\|^2 + 2\|v\|^2. \quad (1)$$

*Proof.* Straightforward computation, convince yourself.  $\square$

**Remark 3.** With La. 3, one can prove that on  $\mathbb{R}^n$  ( $n > 1$ ), all  $p$ -norms, except for  $p = 2$ , are not induced by scalar products. (Exercise!)

Orthogonality plays already an important role in the finite-dimensional case, and so does it in the infinite-dimensional one. In particular, projections onto subspaces are based on orthogonal vectors.

**Definition 7** (Orthogonality and projection). *Let  $\mathcal{X}$  be a Hilbert space, and let  $\mathcal{Y}$  be a subspace of  $\mathcal{X}$ . Two elements  $x, y \in \mathcal{X}$  are said to be orthogonal, if there holds*

$$(x, y) = 0.$$

*The projection of some  $x \in \mathcal{X}$  onto  $\mathcal{Y}$  is defined as the value  $\Pi_{\mathcal{Y}}(x)$  for which*

$$\|x - \Pi_{\mathcal{Y}}(x)\| = \min_{y \in \mathcal{Y}} \|x - y\|.$$

*If the context is clear, we will often write  $\Pi$  instead of  $\Pi_{\mathcal{Y}}$ .*

**Lemma 4.** *If  $\mathcal{Y}$  is non-empty and closed, then  $\Pi_{\mathcal{Y}}(x)$  is well-defined. In particular, the solution of the minimalization problem exists and is unique.*

*Proof.* If  $x \in \mathcal{Y}$ , then  $\Pi_{\mathcal{Y}}(x) = x$ , and there is nothing to prove. We restrict ourselves therefore to the case  $x \notin \mathcal{Y}$ . Let  $\tau := \inf_{y \in \mathcal{Y}} \|x - y\|$ . As  $\mathcal{Y}$  is closed and  $x \notin \mathcal{Y}$ , there must hold  $\tau > 0$ . (Prove this!) Let  $(y_n)_{n \in \mathbb{N}} \subset \mathcal{Y}$  be a sequence such that  $\|y_n - x\| \rightarrow \tau$ . We want to prove that  $y_n$  is a Cauchy sequence. To this end, observe that for two values  $z_1, z_2 \in \mathcal{Y}$ , there holds due to the parallelogram identity (1):

$$\|z_1 - z_2\|^2 = \|(z_1 - x) - (z_2 - x)\|^2 = 2\|z_1 - x\|^2 + 2\|z_2 - x\|^2 - \|(z_1 - x) + (z_2 - x)\|^2.$$

Furthermore,

$$\|(z_1 - x) + (z_2 - x)\|^2 = 4 \left\| \frac{z_1 + z_2}{2} - x \right\|^2 \geq 4\tau^2,$$

because  $\frac{z_1 + z_2}{2} \in \mathcal{Y}$  and therefore, the distance to  $x$  must be  $\geq \tau$ . With that being said, there is

$$\|z_1 - z_2\|^2 \leq 2\|z_1 - x\|^2 + 2\|z_2 - x\|^2 - 4\tau^2. \quad (2)$$



Now returning to the sequence  $(y_n)_{n \in \mathbb{N}}$ . Given an  $\varepsilon$ , choose  $n_0$  such that for all  $n \geq n_0$ , there holds  $|\|y_n - x\| - \tau| \leq \varepsilon$ . Then, for all  $n, m \geq n_0$ , there holds with (2)

$$\|y_n - y_m\| \leq 2\|y_n - x\|^2 + 2\|y_m - x\|^2 - 4\tau^2 \leq 2(\tau + \varepsilon)^2 + 2(\tau + \varepsilon)^2 - 4\tau^2 = 8\tau\varepsilon + 4\varepsilon^2.$$

$(y_n)_{n \in \mathbb{N}}$  is hence a Cauchy sequence. The limit  $\bar{y}$  fulfills  $\|\bar{y} - x\| = \tau$ , we thereby have existence.

Assume now that there are two  $\bar{y}_1, \bar{y}_2 \in \mathcal{Y}$  such that  $\|\bar{y}_i - x\| = \tau$ . Then, again due to (2), there holds

$$\|\bar{y}_1 - \bar{y}_2\|^2 \leq 2\|\bar{y}_1 - x\|^2 + 2\|\bar{y}_2 - x\|^2 - 4\tau^2 = 0,$$

hence  $\bar{y}_1 = \bar{y}_2$ , and hence uniqueness.  $\square$

**Remark 4.** • One can define a projection also for a subset  $\mathcal{Y} \subset \mathcal{X}$  rather than a subspace. Then La. 4 remains true if  $\mathcal{Y}$  is non-empty, closed and convex.

- If  $\mathcal{Y}$  is a subspace, then the projection operator is linear - this is an easy consequence of the following lemma.

**Lemma 5.** Consider the assumptions as in La. 4. Then, there holds:

$$\bar{y} = \Pi_{\mathcal{Y}}(x) \quad \Leftrightarrow \quad (x - \bar{y}, y) = 0, \quad \forall y \in \mathcal{Y}.$$

*Proof.* Without loss of generality, we consider the case that  $\bar{y} = 0$ . This is possible, because of the following: Assume that  $\bar{y} = \Pi_{\mathcal{Y}}(x)$ , i.e.,

$$\inf_{y \in \mathcal{Y}} \|x - y\| = \|x - \bar{y}\|.$$

Furthermore, as  $\bar{y} \in \mathcal{Y}$ , there holds

$$\inf_{y \in \mathcal{Y}} \|x - y\| = \inf_{y \in \mathcal{Y}} \|(x - \bar{y}) - (y - \bar{y})\| = \inf_{y \in \mathcal{Y}} \|(x - \bar{y}) - y\| = \|(x - \bar{y}) - 0\|.$$

Hence,  $\Pi_{\mathcal{Y}}(x - \bar{y}) = 0$ , so we only consider a linear shift.

“ $\Rightarrow$ ”: There holds

$$\|x - \bar{y}\| = \|x - 0\| \leq \|x - y\|, \quad \forall y \in \mathcal{Y}$$

due to the definition of the projection operator. Now fix some  $y \in \mathcal{Y}$ , and consider, for  $t \in \mathbb{R}$ ,  $ty \in \mathcal{Y}$ . Consequently,

$$\|x\|^2 \leq \|x - ty\|^2 = \|x\|^2 + t^2\|y\|^2 - 2t \operatorname{Re}(x, y)$$

and hence

$$2t \operatorname{Re}(x, y) \leq t^2\|y\|^2, \quad \forall t \in \mathbb{R}.$$

Now let first  $t$  be positive, and let  $t \rightarrow 0$ . Thus,  $\operatorname{Re}(x, y) \leq 0$ . For  $t$  negative and  $t \rightarrow 0$ , one gets  $\operatorname{Re}(x, y) \geq 0$ . Thus,

$$\operatorname{Re}(x, y) = 0.$$

In case  $\mathbb{K} = \mathbb{C}$ , note that with  $y \in \mathcal{Y}$  there also holds  $iy \in \mathcal{Y}$  and  $\operatorname{Im}(x, y) = -\operatorname{Re}(x, iy)$ . With the same arguments as before, one gets that also the imaginary part is zero. This concludes the first part.

“ $\Leftarrow$ ”: We can compute:

$$\|x - y\|^2 = \|x\|^2 + \|y\|^2 - 2 \operatorname{Re}(x, y) = \|x\|^2 + \|y\|^2 \geq \|x\|^2 + \|0\|^2,$$

hence  $\Pi_{\mathcal{Y}}(x) = 0$ .  $\square$

**Corollary 2.** Assume that  $\mathcal{Y} = \text{span}(y_i, 1 \leq i \leq N)$  with  $y_i \in \mathcal{X}$ ,  $\mathcal{X}$  Hilbert space,  $N \in \mathbb{N}$ , and  $(y_i, y_j) = 0$  if  $i \neq j$ . Then, there holds

$$\Pi_{\mathcal{Y}}(x) = \sum_{i=1}^N \frac{(x, y_i)}{(y_i, y_i)} y_i.$$

**Example 6.** There is a very close link to Fourier analysis here, see also Def. 6. Remember that

$$L_{\circ}^2 := \{f : \mathbb{R} \rightarrow \mathbb{C} \text{ is measurable} : \|f\|_{L_{\circ}^2} < \infty, f(x + 2\pi) = f(x) \text{ almost everywhere}\}.$$

Periodic functions are best represented as functions 'on a circle', which itself can in the complex plane be described as  $\{e^{it} \mid t \in [0, 2\pi)\}$ . Roughly speaking, it can therefore be said that the role of  $x$  (and  $x^2, x^3, \dots$ ) for non-periodic functions is the same as  $e^{it}$  (and  $e^{2it}, e^{3it}, \dots$ ) for periodic ones. Define therefore  $\mathbf{e}_k(t) := e^{ikt}$  for  $k \in \mathbb{Z}$ , and

$$U_N := \text{span}(\mathbf{e}_{-N}, \dots, 1, \dots, \mathbf{e}_N) \subset L_{\circ}^2.$$

It can be easily shown that  $(\mathbf{e}_k, \mathbf{e}_l)_{L_{\circ}^2} = \delta_{k,l}$  (try it!), and so the  $\mathbf{e}_k$  form an orthonormal basis of the finite-dimensional space  $U_N$ .

The truncated Fourier series  $s_N$  of a function  $f \in L_{\circ}^2$  is defined as the orthogonal projection of  $f$  onto  $U_N$ , hence

$$s_N := \Pi_{U_N}(f). \quad (3)$$

By construction, it is obvious that  $s_N$  is a linear combination of the  $\mathbf{e}_k$ . The corresponding coefficients are denoted by  $\widehat{f}(k)$ , hence

$$s_N =: \sum_{k=-N}^N \widehat{f}(k) \mathbf{e}_k.$$

$\widehat{f}(k)$  are called Fourier coefficients, and can be computed through

$$\widehat{f}(k) = (f, \mathbf{e}_k) \equiv \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{-ikx} dx,$$

see Cor. 2. Also due to construction, and this shows the importance of the truncated Fourier series  $s_N$ , there holds

$$\|f - s_N\|_{L_{\circ}^2} = \min_{s \in U_N} \|f - s\|_{L_{\circ}^2}.$$

## 2 Periodic functions: Fourier series

Before continuing, we remind the reader of the following important property which holds for all  $\varphi \in \mathbb{R}$ :

$$e^{i\varphi} = \cos(\varphi) + i \sin(\varphi).$$

Furthermore, we will make frequent use of

$$\cos(x) = \cos(-x), \quad \sin(x) = -\sin(-x).$$

### 2.1 Fourier transform

We have already seen the definition of the Fourier coefficients  $\hat{f}(k)$  and the truncated Fourier series  $s_N$ . An important guiding question is whether, in what sense, and under what conditions, there holds  $\lim_{N \rightarrow \infty} s_N = f$ . We postpone this discussion for a moment, and start with some simple properties of  $\hat{f}(k)$ :

**Lemma 6.** *The following properties hold:*

- If  $f$  is a real-valued function, then  $\hat{f}(k) = \overline{\hat{f}(-k)}$ .
- If  $f$  is an even function, then  $\hat{f}(k) \in \mathbb{R}$ .
- If  $f$  is an odd function, then  $\hat{f}(k) \in i\mathbb{R}$ .

*Proof.* Prove this! □

**Remark 5.** 'Traditionally', the Fourier series has been introduced as

$$\frac{a_0}{2} + \sum_{k=1}^{\infty} a_k \cos(kx) + b_k \sin(kx)$$

rather than  $\sum_{k \in \mathbb{Z}} \hat{f}(k) e^{ikx}$ . With real  $a_k$  and  $b_k$ , the former is confined to functions  $f : \mathbb{R} \rightarrow \mathbb{R}$ , while the latter also works for functions  $f : \mathbb{R} \rightarrow \mathbb{C}$ . One can observe that there is indeed a simple bijection between the 'traditional' representation and the one with  $\mathbf{e}_k$  involved:

$$\sum_{k=-\infty}^{\infty} \hat{f}(k) \mathbf{e}_k = \hat{f}(0) + \sum_{k=1}^{\infty} \left( (\hat{f}(k) + \hat{f}(-k)) \cos(kx) + (\hat{f}(k) - \hat{f}(-k)) i \sin(kx) \right).$$

Upon defining  $a_k := \hat{f}(k) + \hat{f}(-k)$  and  $b_k := i(\hat{f}(k) - \hat{f}(-k))$ , one observes that there is a simple bijection between the 'traditional' representation and the one with  $\mathbf{e}_k$  involved. Note that because of La. 6,  $a_k$  and  $b_k$  are indeed real if  $f$  is a real-valued function. Stated a bit simpler, one can say that  $a_0 = 2 \operatorname{Re}(\hat{f}(0))$ ; and for  $k > 0$ ,  $a_k = 2 \operatorname{Re}(\hat{f}(k))$  and  $b_k = -2 \operatorname{Im}(\hat{f}(k))$ .

Before starting to prove under what circumstances a Fourier series exists - and converges toward  $f$  - we show some instructive examples.

**Example 7.** •  $f(x) = \sin(x)$ . Obviously, the 'traditional' Fourier transform has  $b_1 = 1$ , and all other parameters are zero. Thus,  $\hat{f}(1) = -\hat{f}(-1) = -\frac{i}{2}$ ,  $\hat{f}(k) = 0$  for  $k \neq \pm 1$ .

- $f(x) = \sin^2(x)$ . Because  $\sin^2(x) = \frac{1}{2} - \frac{\cos(2x)}{2}$  (try it!), the Fourier coefficients are  $\hat{f}(0) = \frac{1}{2}$ ,  $\hat{f}(-2) = \hat{f}(2) = -\frac{1}{4}$ , and  $\hat{f}(k) = 0$  for  $k \notin \{0, \pm 2\}$ .

- Consider the rectangle function  $f(x) = \chi_{(-\frac{1}{2}, \frac{1}{2})}(x)$ . Its Fourier transform is given by  $\widehat{f}(0) = \frac{1}{2\pi}$  and

$$\widehat{f}(k) = \frac{1}{2\pi} \int_{-\frac{1}{2}}^{\frac{1}{2}} e^{-ikx} dx = \frac{-1}{ik2\pi} \left( e^{-i\frac{k}{2}} - e^{i\frac{k}{2}} \right) = \frac{1}{k\pi} \sin\left(\frac{k}{2}\right) =: \frac{1}{2\pi} \operatorname{sinc}\left(\frac{k}{2}\right)$$

The  $\operatorname{sinc}(x) := \frac{\sin(x)}{x}$  (sinus cardinalis) function occurs very often in Fourier analysis, which is why it got an extra name. Note that sometimes, the sinc function also comes in the normalized way, unfortunately, there is no standard in literature. Whenever we talk about the sinc function, we use it in the way presented above.

- $f(x) = \frac{x}{\pi}$  on  $[-\pi, \pi]$ . Note that this function is not smooth! Its Fourier transform can be computed as  $\widehat{f}(0) = 0$  and,  $k \neq 0$ ,

$$\widehat{f}(k) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{x}{\pi} e^{-ikx} dx = \frac{1}{2\pi} \left( \frac{-xe^{-ikx}}{\pi ik} + \frac{e^{-ikx}}{\pi k^2} \right) \Big|_{-\pi}^{\pi} = \frac{e^{-ikx}}{2\pi^2} \left( \frac{ix}{k} + \frac{1}{k^2} \right) \Big|_{-\pi}^{\pi} = \frac{i}{\pi k} \cos(k\pi).$$

So far, we didn't talk about whether the Fourier series exists, and whether it approximates the original function - and in what sense. However, for discontinuous functions, the Fourier series shows a very peculiar behavior at the discontinuity - in particular, the overshoots do not die out as  $N \rightarrow \infty$ . This phenomenon is known as *Gibbs phenomenon* (dutch: *Gibbs-verschijnsel*). Please read about this phenomenon here: [https://en.wikipedia.org/wiki/Gibbs\\_phenomenon](https://en.wikipedia.org/wiki/Gibbs_phenomenon). Read until (including!) "Explanation". In particular, the first three images are important!

We now embark on the journey of proving that  $f \in L^2_{\circ}$  is indeed approximated (in the  $L^2_{\circ}$ -norm) by its Fourier transform. We begin with a direct consequence of the fact that  $s_N$  is the orthogonal projection onto  $U_N$ :

**Theorem 5** (Bessel's (in)equality). *For  $f \in L^2_{\circ}$  and  $s_N$  as defined in (3), there holds:*

$$\|s_N\|_{L^2_{\circ}}^2 + \|f - s_N\|_{L^2_{\circ}}^2 = \|f\|_{L^2_{\circ}}^2. \quad (4)$$

From this follows  $\|s_N\|_{L^2_{\circ}}^2 \leq \|f\|_{L^2_{\circ}}^2$  easily.

*Proof.* This is a special case of Ex. 8! □

The following lemma is one of the most important ones in the context of Fourier analysis. It goes back to the French mathematician Marc-Antoine Parseval:

**Lemma 7** (Parseval's identity). *For any  $f \in L^2_{\circ}$ , there holds:*

$$\|f\|_{L^2_{\circ}}^2 = \sum_{k \in \mathbb{Z}} |\widehat{f}(k)|^2. \quad (5)$$

**Theorem 6.** *With this theorem, one directly obtains for an  $f \in L^2_{\circ}$ :*

$$\|f - s_N\|_{L^2_{\circ}} \rightarrow 0, \quad N \rightarrow \infty. \quad (6)$$

*Proof.* There holds  $\|s_N\|_{L^2_{\circ}}^2 = \sum_{k=-N}^N |\widehat{f}(k)|^2$ . Plugging this into Bessel's equation (4) and exploiting (5) yields the statement. □

*Proof of Parseval's identity, La. 7.* The proof is a bit tedious. W.l.o.g., we only consider functions  $f : \mathbb{R} \rightarrow \mathbb{R}$ , the extension to complex functions is trivial, as any complex function is  $f(x) = \operatorname{Re}(f(x)) + i \operatorname{Im}(f(x))$ .

The proof consists of three steps:

1. We show that the statement holds for a rectangle function

$$f(x) = \begin{cases} 1, & a \leq x \leq b \\ 0, & \text{otherwise} \end{cases}. \quad (7)$$

2. Then, we show that it holds for a finite linear combination of rectangle functions.

3. Once this is done, we know that for any  $f \in L^2_{\circ}$ , one can find a step function  $f_{\varepsilon}$ , such that  $\|f - f_{\varepsilon}\|_{L^2_{\circ}} \leq \varepsilon$ . Let  $s_N$  denote the truncated Fourier series for  $f$ , and  $s_{\varepsilon,N}$  the one for  $f_{\varepsilon}$ . Furthermore, let  $N$  be such that  $\|f_{\varepsilon} - s_{\varepsilon,N}\| \leq \varepsilon$ . (We can choose such an  $N$ , because for step functions, Thm. 6 already holds.) Then, one can conclude:

$$\|f - s_N\|_{L^2_{\circ}} = \|f_{\varepsilon} - s_{\varepsilon,N} + f - f_{\varepsilon} - s_N + s_{\varepsilon,N}\| \leq \|f_{\varepsilon} - s_{\varepsilon,N}\| + \|(f - f_{\varepsilon}) - (s_N - s_{\varepsilon,N})\|.$$

Now  $\|f_{\varepsilon} - s_{\varepsilon,N}\| \leq \varepsilon$  by assumption, and

$$\|(f - f_{\varepsilon}) - (s_N - s_{\varepsilon,N})\|^2 \leq \|f - f_{\varepsilon}\|^2 \leq \varepsilon^2$$

because of (4). Now let  $\varepsilon \rightarrow 0$  and use (4) again.

Now let us work out steps 1 and 2. We remind the reader of the following infinite sums:

$$\sum_{k=1}^{\infty} \frac{1}{k^2} = \frac{\pi^2}{6}, \quad \sum_{k=1}^{\infty} \frac{\cos(kb)}{k^2} = \frac{(b - \pi)^2}{4} - \frac{\pi^2}{12}$$

(For an interesting proof of the first one, look up, e.g., *Basel problem* on the internet. The second one can be found in Analysis 1 by O. Forster.)

1. W.l.o.g. we consider  $a = 0$  and  $0 < b < \pi$ . Then, for  $f$  as in (7), there holds  $\widehat{f}(0) = \frac{b}{2\pi}$ , and for  $k \neq 0$ ,  $\widehat{f}(k) = \frac{1 - e^{-ikb}}{2\pi ik}$ . Summing this up yields:

$$\begin{aligned} \left(\frac{b}{2\pi}\right)^2 + \sum_{k \in \mathbb{Z} \setminus \{0\}} \left| \frac{1 - e^{-ikb}}{2\pi ik} \right|^2 &= \frac{b^2}{(2\pi)^2} + \sum_{k \in \mathbb{Z} \setminus \{0\}} \left( \frac{(1 - \cos(kb))^2 + \sin^2(kb)}{(2\pi)^2 k^2} \right) \\ \frac{b^2}{(2\pi)^2} + \sum_{k \in \mathbb{Z} \setminus \{0\}} \frac{1 - 2\cos(kb) + \cos^2(kb) + \sin^2(kb)}{(2\pi k)^2} &= \frac{b^2}{(2\pi)^2} + \sum_{k \in \mathbb{Z} \setminus \{0\}} \frac{1}{2\pi^2} \frac{1 - \cos(kb)}{k^2} \\ &= \frac{b^2}{(2\pi)^2} + \frac{2}{2\pi^2} \left( \frac{\pi^2}{6} + \frac{\pi^2}{12} - \frac{(b - \pi)^2}{4} \right) = \frac{2b\pi}{4\pi^2} = \frac{b}{2\pi} = \|f\|_{L^2_{\circ}}^2 \end{aligned}$$

2. Now if  $f$  is a finite linear combination of rectangular functions, i.e.,  $f = \sum_{j=1}^m c_j f_j$  with  $f_j$  being a function of form (7), then because of linearity, one can compute

$$\|f - s_N\| = \left\| \sum_{j=1}^m c_j (f_j - s_{j,N}) \right\| \leq \sum_{j=1}^m |c_j| \|f_j - s_{j,N}\|.$$

We know that  $\|f_j - s_{j,N}\|$  converges toward zero, which implicitly proves the statement. □

**Remark 6.** *There is a word of warning necessary here: Convergence in  $L^2$  and pointwise convergence is not the same! Let us clarify when one follows from the other:*

- *If the domain  $\Omega$  where the function is defined is bounded, and some sequence of functions  $f_N$  converges uniformly toward  $f$ , then it also converges in  $L^2$ , as one can compute*

$$2\pi \|f - f_N\|_{L^2_{\circ}}^2 = \int_{\Omega} |f(x) - f_N(x)|^2 dx \leq |\Omega| \|f(x) - f_N(x)\|_{\infty}^2 \rightarrow 0.$$

- All other implications do in general not hold. As an example, let  $f_N(x) := N \cdot \chi_{(0, \frac{1}{N})}(x)$ . Obviously,  $f_N$  converges pointwise toward  $f \equiv 0$ , however, not uniformly. Furthermore,

$$2\pi \|f - f_N\|_{L^2_0}^2 = N \rightarrow \infty.$$

- However, there is the following statement: If a sequence of functions  $(f_N)_{N \in \mathbb{N}}$  converges in  $L^2$  to a function  $f$ , then one can extract a subsequence that converges pointwise almost everywhere.

**Lemma 8** (Riemann-Lebesgue-Lemma). For  $f \in L^2_0$ , there holds  $\widehat{f}(k) \rightarrow 0$  for  $k \rightarrow \pm\infty$ .

*Proof.* As the  $\mathbf{e}_k$  form an orthonormal basis, there holds  $\|s_N\|_{L^2_0}^2 = \sum_{k=-N}^N |\widehat{f}(k)|^2$ . From Bessel's inequality one can directly conclude that  $\widehat{f}(k) \rightarrow 0$  for  $k \rightarrow \pm\infty$ , because otherwise, the infinite sum diverges.  $\square$

It is the more fascinating that there hold very general statements for the pointwise convergence for Fourier series. Without a proof, we start with the following important theorem, proved by *Carleson* in 1966:

**Theorem 7.** Let  $f \in L^2_0$ . Then,  $s_N(x)$  converges almost everywhere toward  $f(x)$ , i.e., there holds  $\lim_{N \rightarrow \infty} s_N(x) = f(x)$  for almost every  $x \in (-\pi, \pi)$ .

This is less precise than what one aims at. Uniform convergence can be achieved if the function is of *bounded variation*:

**Definition 8.** Let an arbitrary subdivision of  $(-\pi, \pi)$  be given as

$$\mathcal{T} : -\pi = t_0 < t_1 < \dots < t_n = \pi.$$

If there holds

$$\sup_{\mathcal{T}} \sum_{j=1}^n |f(t_j) - f(t_{j-1})| < \infty$$

then  $f$  is said to be of *bounded variation*. The supremum is called the (total) variation  $V(f)$  of  $f$ .

To get an idea of the concept of variation, we state the following lemma:

**Lemma 9.** Let  $f \in C^1$ . Then,

$$V(f) = \int_{-\pi}^{\pi} |f'(x)| dx.$$

*Proof.* Exercise. Exploit that  $f(t_k) - f(t_{k-1}) = f'(\xi)(t_k - t_{k-1})$ , where  $\xi \in (t_{k-1}, t_k)$ .  $\square$

However, not only differentiable functions have a variation. As an example, the function  $\chi_{(-\pi, 0)}$  has variation of two. (Convince yourself!)

With this being said, one can state the following Theorem:

**Theorem 8.** Let  $f \in L^2_0$  be of bounded variation. Then, for any  $x$ , there holds:

$$s_N(x) \rightarrow \frac{f(x^+) + f(x^-)}{2}.$$

If  $f$  is, in addition, continuous, then convergence is uniform.

*Proof.* The proof needs much more preparation than we have the time fore. A core ingredient, however, is the decomposition of a bounded-variation function  $f$  into the difference of two monotonically increasing functions  $g_1$  and  $g_2$ . (This is called Jordan decomposition.)  $\square$