# MATH5111S24: $L^p$ Spaces (Prep for 03/18 & 03/20 Lectures)

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## 1 Intro

Our starting point is a measure space  $(\Omega, \mathcal{F}, \mu)$ . As usual, we will assume this measure space is complete and to avoid trivialities, we will assume that there exists  $A \in \mathcal{F}$  with  $\mu(A) \in (0, \infty)$ .

We proved that the sum of two measurable functions is measurable and that a constant times a measurable function is measurable. Therefore the set  $\mathcal{L}$  of real-valued measurable functions is a vector space with respect to pointwise addition and scalar multiplication. Note that this structure has nothing to do with the measure, just the sigma-algebra.

For  $f \in \mathcal{L}$  and  $p \in [1, \infty)$ . Define

$$||f||_p = (\int |f|^p d\mu)^{\frac{1}{p}},$$

and

$$\mathcal{L}^p = \{ f \in \mathcal{L} : \|f\|_p < \infty \}.$$

In what follows we will always write q for the conjugate exponent to p, defined through the relation:

$$\frac{1}{p} + \frac{1}{q} = 1.$$
 (1)

For example if p = 2, q = 2 and if p = 3,  $q = \frac{3}{2}$ , etc.

Note that  $q \in (1, \infty]$ , and that  $q = \infty$  if and only if p = 1. The fact that the conjugate of p = 1 is  $q = \infty$  suggests (as will become apparent through Holder's inequality, Theorem 2.2) that we may want to introduce and define  $\mathcal{L}^{\infty}$ . We will do that in Section 4.

The following identity is a restatement of the relation (1):

$$p-1 = \frac{p}{q}.$$

## 2 Two Inequalities: Holder & Minkowski

Holder's inequality is among the most important inequalities in analysis. It is a generalization of the Cauchy-schwarz inequality. It is derived from one of the oldest tricks in the playbook, the Arithmetic-Geometric inequality.

**Proposition 2.1** (Arithmetic-Geometric Inequality). Let  $\alpha, \beta$  be nonnegative and  $\lambda \in (0, 1)$ . Then

$$\alpha^{\lambda}\beta^{1-\lambda} \le \lambda\alpha + (1-\lambda)\beta$$

with a strict inequality if and only if  $\alpha \neq \beta$ .

The proof is equivalent to the strict convexity of the exponential function. Indeed if  $\alpha\beta = 0$  the inequality is trivial and otherwise the lefthand side is  $e^{\lambda \ln \alpha + (1-\lambda) \ln \beta}$ , and the strict convexity of the exponential implies this is  $\leq \lambda e^{\ln \alpha} + (1-\lambda)e^{\ln \beta}$ , with equality if and only if  $\ln \alpha = \ln \beta$ .

**Theorem 2.2** (Holder's Inequality). Let  $p \in (1, \infty)$  and let q be its conjugate from (1). Let f and g be measurable and nonnegative. Then

$$\int fgd\mu \le \|f\|_p \|g\|_q,$$

where the righthand side is defined as 0 if one of the factors is = 0. If  $f \in \mathcal{L}^p$ and  $g \in \mathcal{L}^q$ , then an equality holds if and only f = 0  $\mu$ -a.e., g = 0  $\mu$ -a.e. or there exists some C > 0 such that  $f^p = Cg^q \mu$ -a.e.

We note that the condition for equality can be restated as  $f^p$  and  $g^q$  are linearly dependent,  $\mu$ -a.e.

*Proof.* We observe that if  $||f||_p = 0$  then f = 0  $\mu$ -a.e., and consequently fg = 0  $\mu$ -a.e., and therefore both the lefthand side and the righthand side are zero. The same holds if  $||g||_q = 0$ .

We can therefore assume that both  $||f||_p$  and  $||g||_q$  are strictly positive. If any of these is infinite, then the inequality is trivial.

We are left with the case  $||f||_p$ ,  $||g||_q \in (0, \infty)$ . We will make yet another reduction which will save some work. Let  $F = f/||f||_p$  and  $G = g/||g||_q$ . Then

$$\int fgd\mu = \|f\|_p \|g\|_q \int FGd\mu$$

Therefore Holder's inequality is equivalent to

$$\int FGd\mu \le 1 \tag{2}$$

We note that the definition of F and G,  $||F||_p = ||G||_q = 1$ . Let's now use the AGM, Proposition 2.1, with  $\lambda = \frac{1}{p}$  and therefore  $1 - \lambda = \frac{1}{q}$ ,  $\alpha = F^p(\omega)$  and  $\beta = G^p(\omega)$ , which then gives

$$(FG)(\omega) \le \frac{1}{p} F^p(\omega) + \frac{1}{q} G^q(\omega).$$
(3)

with a strict inequality on the set  $A = \{\omega : F^p(\omega) \neq G^q(\omega)\}$ . Now integrate both sides of (3) to obtain (2). An equality holds if and only  $\mu(A) = 0$ , equivalently

 $A^c \ \mu$ -a.e. Now  $A^c = \{f^p = \frac{\|f\|_p^p}{\|g\|_q^q} g^q\}$ , which implies that for some C > 0,  $f^p = Cg^q$ ,  $\mu$ -a.e. Integrating the last equality gives  $\|f\|_p^p = C\|g\|_q^q$ , or  $C = \|f\|_p^p / \|g\|_q^q$ , that is C is determined, and so that if for some C > 0,  $f^p = Cg^q \ \mu$ -a.e., then  $A^c \ \mu$ -a.e., and so an equality holds in Holder's inequality.

A very important corollary to Holder's inequality is the following, the triangle inequality for  $\|\cdot\|_p$ :

**Theorem 2.3** (Minkowski's Inequality). Let  $p \in (1, \infty)$  and let  $f, g \in \mathcal{L}^p$ . Then  $\|f + g\|_p \leq \|f\|_p + \|g\|_p$ .

Clearly if  $f \in \mathcal{L}^p$  and  $c \in \mathbb{R}$ , then  $cf \in \mathcal{L}^p$ . Minkowski's inequality implies that if  $f, g \in \mathcal{L}^p$ , then  $f + g \in \mathcal{L}^p$ , and so we obtain the following:

**Corollary 2.4.**  $\mathcal{L}^p$  is a vector space with respect to pointwise addition and scalar multiplication.

*Proof.* If the lefthand side is zero there is nothing to prove. We will proceed assuming  $||f + g||_p > 0$  (apriori it may be infinite). To get the lefthand side we need to integrate the function  $|f + g|^p$ . Use the triangle inequality to obtain

$$|f+g|^p = |f+g||f+g|^{p-1} \le (|f|+|g|)|f+g|^{p-1}.$$
(4)

Now  $|f + g| \le 2 \max(|f|, |g|)$ , and so

$$|f+g|^{p-1} \le 2^{p-1} \max(|f|^{p-1}, |g|^{p-1}).$$

Recalling that  $p-1 = \frac{p}{q}$ , it follows that

$$(|f+g|^{p-1})^q \le 2^p \max(|f|^p, |g|^p) \le 2^p (|f|^p + |g|^p).$$

Since  $f, g \in \mathcal{L}^p$ ,  $|f + g|^{p-1} \in \mathcal{L}^q$ . Now,

$$|||f+g|^{p-1}||_q^q = \int (|f+g|^{p-1})^q = \int |f+g|^p d\mu = ||f+g||_p^p,$$

that is

$$\|f+g\|^{p-1}\|_{q} = \|f+g\|_{p}^{p/q}.$$
(5)

Now integrate both sides of (4)

$$\begin{split} \|f+g\|_{p}^{p} &= \int |f+g|^{p} d\mu \\ &\leq \int |f| |f+g|^{p-1} d\mu + \int |g| |f+g|^{p-1} d\mu \\ &\leq \|f\|_{p} \|f+g|_{p}^{p/q} + \|g\|_{p} \|f+g\|_{p}^{p/q}, \end{split}$$

where the last line was obtained from Holder's inequality and (5). Divide by sides by  $||f + g||_p^{p/q} = ||f + g||_p^{p-1}$  to obtain the result.

## 3 $L^p$ , a Complete Normed Space

#### 3.1 Normed Spaces

Recall the following definition:

**Definition 1.** A semi-norm  $\|\cdot\|$  on a vector space V is a mapping  $\|\cdot\| \to [0,\infty)$  with the following properties

- 1. (homogeneity) ||cv|| = |c|||v||, for all  $c \in \mathbb{R}$ , and  $v \in V$ .
- 2. (triangle inequality)  $||v + u|| \le ||v|| + ||u||$  for all  $v, u \in V$ .

A semi-norm is a norm if in addition it satisfies

0. ||v|| = 0 if and only if v = 0.

If  $\|\cdot\|$  is a norm, then we call  $(V, \|\cdot\|)$  a normed space.

Now  $\|\cdot\|_p$  is a semi-norm on the vector space  $\mathcal{L}^p$ . In general, it is not a norm. Consider the Lebesgue measure:  $\|\mathbf{1}_{\mathbb{Q}}\|_p = \|0\|_p$ , though these two functions are distinct elements of  $\mathcal{L}$ . We remedy this through a simple construction of a quotient space.

Suppose that V is a vector space and  $\|\cdot\|: V \to [0, \infty)$  is a semi-norm on V. Let  $V_0 = \{v \in V : \|v\| = 0\}$ . Note that  $V_0$  is automatically a subspace of V because of the semi-norm properties.

For every  $v \in V$ , let  $v + V_0 = \{v + v_0 : v_0 \in V_0\}$ . Then for  $v, v' \in V$  either  $v - v' \in V_0$ , in which case  $v + V_0 = v' + V_0$  or  $v - v' \notin V_0$ , in which case  $(v + V_0) \cap (v' + V_0) = \emptyset$  (check!). Let  $V/V_0$  be the set  $\{v + V_0 : v \in V\}$  (in other words: the relation  $v - v' \in V_0$  is an equivalence relation on V, and  $V/V_0$  is the set of equivalence classes). We call  $V/V_0$  the quotient of V over  $V_0$ .

We can equip the quotient space  $V/V_0$  with a vector space structure in the most obvious way: the addition of  $(v + V_0) + (u + V_0)$  is defined as  $(u + v) + V_0$  and the scalar multiplication  $c(v + V_0)$  is defined as  $(cv) + V_0$ . The zero element in this vector space,  $0_{V/V_0}$ , is  $0 + V_0$ . Abusing notation, extend  $\|\cdot\|$  to  $V/V_0$  by letting  $\|v + V_0\| = \|v\|$ . We will show that this mapping is well defined and is a norm.

First, let's show it is well-defined. Suppose  $v + V_0 = v' + V_0$ , then  $||v|| = ||v'| + \underbrace{(v - v')}_{\in V_0}|| \le ||v'|| + ||v - v'|| = ||v'||$ , due to the triangle inequality. As

the inequality holds with the roles of v and v' interchanged, it follows that ||v'|| = ||v|| and therefore the mapping is well-defined.

It immediately follows from the definition that  $\|\cdot\|$  is a semi-norm on  $V/V_0$ . To show it is a norm, observe that  $\|v + V_0\| = 0$  if and only if  $\|v\| = 0$  if and only if  $v \in V_0$  if and only if  $v + V_0 = 0_{V/V_0}$ . Therefore  $(V/V_0, \|\cdot\|)$  is a normed space.

Going back to our main topic. The mapping  $\|\cdot\|_p$  is a semi-norm on  $\mathcal{L}^p$ . Indeed, from the definiton, if  $f \in \mathcal{L}^p$  and  $c \in \mathbb{R}$ , then  $\|cf\|_p = \|c\|\|f\|_p$ . The triangle inequality for  $\|\cdot\|_p$  is Minkowski's inequality, Theorem 2.3. The normed space  $(\mathcal{L}^p/\mathcal{L}^p_0, \|\cdot\|_p)$  obtained through the construction above is called  $L^p$ . Here  $\mathcal{L}^p_0 = \{f \in \mathcal{L}^p : \|f\|_p = 0\}$ , namely, all functions in  $\mathcal{L}^p$  (equivalently,  $\mathcal{L}$ ), which are 0  $\mu$ -a.e. An element in  $L^p$  is a set of the form  $\{f + h : h = 0, \mu - \text{ a.e.}\}$ , namely all functions equal to  $f \mu$ -a.e. It would be convenient to denote this element by [f].

We turn to a very important property of our newly minted normed space, completeness.

**Definition 2.** Let  $(V, \|\cdot\|)$  be a normed space.

- 1. A sequence  $(v_n : n \in \mathbb{N})$  in V is convergent if there exists  $v \in V$  such that  $\lim_{n\to\infty} ||v_n v|| = 0$ , in which case we say that the sequence has a limit v or that the sequence converges to v, denoted by  $\lim_{n\to\infty} v_n = v$ .
- 2. A sequence  $(v_n : n \in \mathbb{N})$  is a Cauchy sequence if for every  $\epsilon > 0$  there exists  $N = N(\epsilon)$  such that  $||v_n v_{n'}|| \le \epsilon$  for all  $n, n' \ge N$ . Equivalently,  $\lim_{n\to\infty} \sup_{m\in\mathbb{N}} ||v_{n+m} v_n|| = 0$ .
- 3.  $(V, \|\cdot\|)$  is complete if every Cauchy sequence is convergent.

Note that it is very easy to see that every convergent sequence is Cauchy. Yet not every normed space is complete. For example,  $\mathbb{R}^d$  with the Euclidean norm  $||(x_1, \ldots, x_d)|| = \sqrt{x_1^2 + \cdots + x_d^2}$  is complete, yet  $\mathbb{Q}^d$  with the same norm is clearly not complete. Note that the former is  $\mathcal{L}^2$  with  $\mu$  being the counting measure on  $\{1, \ldots, d\}$ .

We also note that it immediately follows from the definition that every Cauchy sequence is bounded in the following sense. If  $(v_n : n \in \mathbb{N})$  is Cauchy, then  $\sup_n ||v_n|| < \infty$ . Indeed, pick  $n_1$  such that  $\sup_m ||v_{n_1+m} - v_{n_1}| \le 1$ . Then for  $n \le n_1$ ,  $||v_n|| \le \max(||v_1||, \ldots, ||v_{n_1}||)$  and for  $n > n_1$ ,  $||v_n|| \le ||v_{n_1}|| + 1$ .

#### **3.2** Completeness of $L^p$

In the last section we constructed a normed vector space  $(L^p, \|\cdot\|_p)$ . We briefly describe its structure. For every  $f \in \mathcal{L}^p$ , let [f] denote all functions in  $\mathcal{L}^p$  (or more generally  $\mathcal{L}$ ) which are equal to  $f \mu$ -a.e. Each of these sets is an element in  $L^p$ . Addition in  $L^p$  and scalar multiplication are defined by the rules [f + g] = [f] + [g] and c[f] = [cf]. The norm of [f] is  $\|f\|_p$ . We prove the following:

**Theorem 3.1.**  $(L^p, \|\cdot\|_p)$  is a complete metric space.

*Proof.* We only need to prove completeness.

**1. Prep.** Let  $([f_n] : n \in \mathbb{N})$  be a sequence in  $L^p$ . Clearly, there exists  $[f] \in L^p$  such that  $\lim_{n\to\infty} [f_n] = [f]$  if and only if  $\lim_{n\to\infty} ||[f_n] - [f]||_p = 0$ . From the definition of the norm  $|| \cdot ||_p$  on  $L^p$ , the latter holds if and only if  $\lim_{n\to\infty} ||f_n - f||_p = 0$  (of course we could take instead any  $f'_n \in [f_n], f' \in [f]$ ).

2. Candidate for f. Yes, we find a candidate for f. Take any subsequence  $(n_k : k \in \mathbb{N})$  tending to infinity. Then

$$f_{n_k} = f_{n_1} + \sum_{l=1}^{k-1} (f_{n_{l+1}} - f_{n_l}).$$
(6)

Therefore,

$$|f_{n_k}| \le |f_{n_1}| + \sum_{l=1}^{k-1} |f_{n_{l+1}} - f_{n_l}|.$$

It follows from Minkowski's inequality, Theorem 2.3, that

$$||f_{n_k}||_p \le ||f_{n_1}||_p + \sum_{l=1}^{k-1} ||f_{n_{l+1}} - f_{n_l}||_p$$

That's true for any subsequence. We now pick a subsequence so that  $(f_{n_l} : l \in \mathbb{N})$ converges  $\mu$ -a.e. Pick  $n_1 = \min\{n : \sup_m \|f_{n+m} - f_n\|_p < 4\}$ , and continue inductively, letting  $n_{l+1} = \min\{n > n_l : \sup_m \|f_{n+m} - f_n\|_p < 4^{-(l+1)}\}$ . This is possible due to the definition of a Cauchy sequence. Let  $A_{l+1} = \{|f_{n_{l+1}} - f_{n_l}| \ge 2^{-(l+1)}) \le 2^{l+1}\}$ . Now by Markov's inequality,

$$\mu(A_{l+1}) \le 2^{l+1} \|f_{n_{l+1}} - f_{n_l}\|_p \le 2^{l+1} 4^{-(l+1)} = 2^{-(l+1)}.$$

Therefore, the series  $\sum \mu(A_l)$  converges, and in particular  $\lim_{n\to\infty} \sum_{l\geq n} \mu(A_l) =$ 0.

$$\mu(\limsup A_l) = \mu(\bigcap_{n=1}^{\infty} \bigcup_{l \ge n} A_l) \le \mu(\bigcup_{l \ge n} A_l) \le \sum_{l \ge n} \mu(A_l) = 0$$

In other words, for all but finitely many *l*'s,  $|f_{n_{l+1}} - f_{n_l}| < 2^{-(l+1)}$ ,  $\mu$ -a.e. In particular,  $\sum_{l=1}^{\infty} |f_{n_{l+1}} - f_{n_l}|$  converges  $\mu$ - a.e. or, the series whose partial sums appear in (6) converges absolutely,  $\mu$ -a.e. As a result,  $\lim_{k\to\infty} f_{n_k}$  converges  $\mu$ -a.e. Denote its limit by f.

3. Candidate in  $\mathcal{L}^p$ . This is basically Fatou's lemma which states:

$$\liminf \int |f_{n_k}|^p d\mu \ge \int \liminf |f_{n_k}|^p d\mu = \int |f|^p d\mu.$$

As a result,  $\liminf ||f_{n_k}||_p \ge ||f||_p$ . The lefthand side is finite because our sequence is Cauchy hence bounded.

4. Convergence of subsequence in  $L^p$ . Fix some k. Then repeating the argiment from the previous step,

$$\liminf_{l\to\infty}\int |f_{n_l}-f_{n_k}|^pd\mu\geq\int |f_{n_k}-f|^pd\mu.$$

That is  $\liminf_{l\to\infty} \|f_{n_k} - f_{n_l}\|_p \ge \|f_{n_k} - f\|_p$ . Therefore,

$$\sup_{n \in \mathbb{N}} \|f_{n_k} - f_{n_k + m}\|_p \ge \|f_{n_k} - f\|_p$$

As  $k \to \infty$  the lefthand side tends to 0 as our original sequence is Cauchy, and therefore  $\lim_{k\to\infty} ||f_{n_k} - f||_p = 0$ .

5. Convergence of full sequence. We have finally arrived at our last step. Sit back and relax. It's all triangle inequality. For every  $n \ge n_1$  there exists a unique k such that  $n_k \le n < n_{k+1}$ . Now

$$||f_n - f||_p \le ||f_n - f_{n_k}||_p + ||f_{n_k} - f||_p.$$

As  $n \to \infty$ ,  $k \to \infty$ . Therefore, the first summand on the righthand side tends to 0 because our sequence is Cauchy. The second also tends to 0 because of the previous step. Done.

## 4 $L^{\infty}$

In this section we complete the description of the  $L^p$  spaces by introducing the space  $L^{\infty}$ . We begin with some motivation. A simple calculus exercise shows that if  $a_1, \ldots, a_d$  are real numbers then  $\lim_{p\to\infty} (\sum_{n=1}^d |a_n|^p)^{1/p} = \max_{n=1,\ldots,d} |a_n|$ . If we equip the finite set  $\{1,\ldots,d\}$  with the counting norm, then the lefthand side can be viewed as the limit of the  $L^p$ -norm of the function  $n \to a_n$  as  $p \to \infty$ . The normed space  $(L^{\infty}, \|\cdot\|_{\infty})$  will be a generalization of this maximum.

For  $f \in \mathcal{L}$ , let

$$||f||_{\infty} = \inf\{L : \mu(|f| > L) = 0\}.$$

Of course,  $||f||_{\infty} \leq \sup |f|$ . A good example to remember is one we have seen before. Consider a Lebesgue measure. Then  $||\mathbf{1}_{\mathbb{Q}}||_{\infty} = 0 < 1 = \sup |\mathbf{1}_{\mathbb{Q}}|$ . As before we define  $\mathcal{L}^{\infty}$  as  $\{f \in \mathcal{L} : ||f||_{\infty} < \infty\}$ .

 $\mathcal{L}^{\infty}$  is a vector space with respect to addition and scalar multiplication of functions and  $\|\cdot\|_{\infty}$  is a semi-norm. The proofs are much simpler than for  $\|\cdot\|_p$ , where we had to get through Holder's inequality to obtain the triangle inequality, Minkowski's inequality. Let's show the triangle inequality for  $\|\cdot\|_{\infty}$ . Let  $f, g \in \mathcal{L}^{\infty}$ , and let  $M_f$  and  $M_g$  be any real numbers strictly larger than  $\|f\|_{\infty}$  and  $\|g\|_{\infty}$ , respectively. Then  $\mu(|f| > M_f) = 0$  and  $\mu(|g| > M_g) = 0$ . If  $|f+g| > M_f + M_g$ , then  $|f| + |g| > M_f + M_g$ , which implies  $|f| > M_f$  or  $|g| > M_g$ . Therefore the set  $\{|f+g| > M_f + M_g\}$  is contained in the set of measure zero  $\{f > M_f\} \cup \{g > M_g\}$ . This implies

$$\|f+g\|_{\infty} \le M_f + M_g.$$

Taking the infimum over allowed values of  $M_f$  and  $M_g$  and using the definition of  $\|\cdot\|_{\infty}$  then gives

$$||f + g||_{\infty} \le ||f||_{\infty} + ||g||_{\infty}.$$

Functions in  $\mathcal{L}^{\infty}$  are also called essentially bounded (with respect to the given measure): bounded, with the exception of a set of measure zero. Consider again the Lebesgue measure. Let  $f(x) = \frac{1}{|x|}$  if x is nonzero rational and 0 otherwise. Then  $||f||_{\infty} = 0$ , although f is unbounded. It is essentially bounded.

We note that for  $f \in \mathcal{L}^{\infty}$ ,  $|f| \leq ||f||_{\infty} \mu$ -a.e.

Repeating the construction of  $L^p$  we obtain the normed space  $(L^{\infty}, \|\cdot\|_{\infty})$ where each element in  $L^{\infty}$  is of the form  $\{f + h : f \in \mathcal{L}^{\infty}, h = 0, \mu \text{ a.e.}\}$ , a set we denote by [f], as usual.

Next we want to prove that  $(L^{\infty}, \|\cdot\|_{\infty})$  is complete. This is much easier than for  $L^p$ , as convergence in this space is uniform convergence, except on a set of measure zero. Indeed, let  $([f_n]: n \in \mathbb{N})$  be a Cauchy sequence. Let  $A_n = \{|f| > \|f\|_{\infty}\}$  and let  $A = \bigcup_{n=1}^{\infty} A_n$ . Then  $\mu(A) = 0$  and on  $A^c$ ,  $|f_n| \le \|f_n\|_{\infty}$ . In particular for every  $\omega \in A^c$  and every  $m, n \in \mathbb{N}$  we have that

$$|f_n - f_m|(\omega) \le ||f_n - f_m||_{\infty}.$$

Therefore, for  $\omega \in A^c$ , the numerical sequence  $(f_n(\omega) : n \in \mathbb{N})$  is Cauchy (in  $\mathbb{R}$ ) and therefore converges to some limit  $f(\omega)$ . Moreover for all  $\omega \in A^c$ ,

$$|f_n - f| = \lim_{m \to \infty} |f_n - f_{n+m}| \le \sup_{m \in \mathbb{N}} ||f_n - f_{n+m}||_{\infty},$$

therefore the convergence is uniform on  $A^c$ , and since  $\mu(A) = 0$ , this implies  $\lim_{n\to\infty} ||f_n - f||_{\infty} = 0$ .

The last result we would like to prove is Holder's inequality. This is even simpler. Let  $f \in \mathcal{L}^1$  and  $g \in \mathcal{L}^\infty$ . Then

$$|\int fgd\mu| \leq \int |f| ||g||_{\infty} d\mu = ||f||_1 ||g||_{\infty}.$$