

# Lecture 15: $L_p$ Spaces and Inequalities

## $L_p$ Spaces

**Definition 1** ( $L_p$  Space). Let  $p \geq 1$ . The  $L_p$  space is defined as:

$$L_p = \{X : \Omega \rightarrow \mathbb{R} : E[|X|^p] < \infty\}$$

$L_p$  is a linear space / vector space.

If  $X_n \in L_p$ , then we say  $X_n \xrightarrow{L_p} X$ , where  $X \in L_p$ , if  $E[|X_n - X|^p] \rightarrow 0$  as  $n \rightarrow \infty$ .

## The $L_p$ Norm

We define a 'norm'  $\|\cdot\|_p : L_p \rightarrow [0, \infty)$

$$X \mapsto \|X\|_p = (E[|X|^p])^{\frac{1}{p}}$$

Properties:

1.  $\|X\|_p \geq 0$
2.  $\|c \cdot X\|_p = |c| \cdot \|X\|_p$
3.  $\|X\|_p = 0 \iff X = 0$  a.s. [Warning : It is not  $X=0$ ]

## Triangle Inequality

Is the following true?

$$\|X + Y\|_p \leq \|X\|_p + \|Y\|_p$$

## Inequalities

### ① Holder's Inequality

Let  $p \geq 1$  and  $q \geq 1$  s.t.  $\frac{1}{p} + \frac{1}{q} = 1$

Holder pair (p,q) conjugate $q = \frac{p}{p-1}$	$(p, q)$	Note: $p = 2 \rightarrow L_2$ $q = 2$ unique s.t. when $p = q$ .
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Then,  $E[|X \cdot Y|] \leq \|X\|_p \cdot \|Y\|_q$

Proof take  $p = q = 2$ .

$$\begin{aligned} \text{corr. } |E[X \cdot Y]| &\leq s.E(X) \cdot s.E(Y) \\ |E[X \cdot Y]| &\leq \sqrt{E[X^2]E[Y^2]} \end{aligned}$$

Cauchy Schwarz Inequality

**Inequality 1** (Minkowski's Inequality). *Let  $p \geq 1$ , then*

$$\|X + Y\|_p \leq \|X\|_p + \|Y\|_p$$

**Corollary 1.** *The triangle inequality for  $L_p$  ( $\|\cdot\|_p$ ) holds whenever  $p \geq 1$ .*

## Proof of Minkowski's Inequality

**Step 1:** There is nothing to prove if  $\|X\|_p = \infty$  or  $\|Y\|_p = \infty$ .

**Step 2:** It is enough to prove the case when  $\|X\|_p < \infty$  and  $\|Y\|_p < \infty$ . Note that if  $X, Y \in L_p$ , then  $\|X + Y\|_p < \infty$ .

Furthermore, without loss we can assume  $X, Y \geq 0$ . Why? Because  $\|X + Y\|_p = (E[|X + Y|^p])^{\frac{1}{p}}$ , and we know  $\|X\|_p$  are dealt with absolute values. If we can prove the inequality for the non-negative random variables, consider  $|X|$  and  $|Y|$  etc:

$$\||X| + |Y|\| \leq \||X|\| + \||Y|\|$$

so we have to just show:

$$(E[(|X| + |Y|)^p])^{\frac{1}{p}} \leq (E[|X|^p])^{\frac{1}{p}} + (E[|Y|^p])^{\frac{1}{p}}$$

Then, since  $E[|X + Y|^p] \leq E[(|X| + |Y|)^p]$ , the result follows.

So, let's assume  $X, Y \geq 0$  and prove:

$$(E[(X + Y)^p])^{\frac{1}{p}} \leq (E[X^p])^{\frac{1}{p}} + (E[Y^p])^{\frac{1}{p}}$$

Consider  $E[(X + Y)^p]$ :

$$\begin{aligned} E[(X + Y)^p] &= E[(X + Y)(X + Y)^{p-1}] \\ &= E[X(X + Y)^{p-1} + Y(X + Y)^{p-1}] \\ &= E[X(X + Y)^{p-1}] + E[Y(X + Y)^{p-1}] \end{aligned}$$

Apply Hölder's Inequality to each term. Let  $V = (X + Y)^{p-1}$ . We need the conjugate exponent  $q$  for  $p$ :  $\frac{1}{p} + \frac{1}{q} = 1 \implies q = \frac{p}{p-1}$ . Note that  $(p-1)q = p$ .

$$\begin{aligned} E[X \cdot V] &\leq \|X\|_p \cdot \|V\|_q \\ E[Y \cdot V] &\leq \|Y\|_p \cdot \|V\|_q \end{aligned}$$

Summing them:

$$\begin{aligned} E[(X + Y)^p] &\leq \|X\|_p \cdot \|V\|_q + \|Y\|_p \cdot \|V\|_q \\ E[(X + Y)^p] &\leq (\|X\|_p + \|Y\|_p) \cdot \|V\|_q \end{aligned}$$

Now, let's substitute  $\|V\|_q$ :

$$\|V\|_q = (E[|V|^q])^{\frac{1}{q}} = (E[((X + Y)^{p-1})^q])^{\frac{1}{q}} = (E[(X + Y)^p])^{\frac{1}{q}}$$

So,

$$E[(X + Y)^p] \leq (\|X\|_p + \|Y\|_p) (E[(X + Y)^p])^{\frac{1}{q}}$$

Since  $\frac{1}{q} = 1 - \frac{1}{p}$ :

$$E[(X + Y)^p] \leq (\|X\|_p + \|Y\|_p) (E[(X + Y)^p])^{1 - \frac{1}{p}}$$

If  $E[(X + Y)^p] = 0$ , the inequality holds. If not, we can divide:

$$\begin{aligned} \frac{E[(X + Y)^p]}{(E[(X + Y)^p])^{1 - \frac{1}{p}}} &\leq \|X\|_p + \|Y\|_p \\ (E[(X + Y)^p])^{1 - (1 - \frac{1}{p})} &\leq \|X\|_p + \|Y\|_p \\ (E[(X + Y)^p])^{\frac{1}{p}} &\leq \|X\|_p + \|Y\|_p \\ \|X + Y\|_p &\leq \|X\|_p + \|Y\|_p \quad \square \end{aligned}$$

## Proof of Hölder's Inequality

To show: Take  $p \geq 1$ ,

$E[|XY|] \leq \|X\|_p \cdot \|Y\|_q$  where  $\frac{1}{p} + \frac{1}{q} = 1$ .

Let  $A = \|X\|_p$  and  $B = \|Y\|_q$ .

**Claim:** WLOG, we can assume  $0 < A, B < \infty$ . Why?

- If  $A = 0$ , then  $X = 0$  a.s., so  $XY = 0$  a.s., and similarly  $B$  is also trivial.
- If  $A = \infty$  or  $B = \infty$ , the RHS is  $\infty$ , and the inequality holds trivially.

Without loss we can assume  $X, Y \geq 0$ , why?

We can just use  $|X|, |Y|$  otherwise. Note  $|XY| = |X| \cdot |Y|$

To show  $E[X \cdot Y] \leq (E[X^p])^{\frac{1}{p}} \cdot (E[Y^q])^{\frac{1}{q}}$  when  $p \geq 1$ , &  $\frac{1}{p} + \frac{1}{q} = 1$ .

& furthermore  $0 < E[X^p], E[Y^q] < \infty$

**Lemma 1** (From Analysis).  $p \geq 1, a, b \geq 0$

(Digression: Convex Function) A function  $f: \mathbb{R} \rightarrow \mathbb{R}$  is called **CONVEX** if  $\forall x_1, x_2$  and  $\alpha \in [0, 1]$ ,  $\forall x_2 \geq x_1$ :

$$f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2)$$