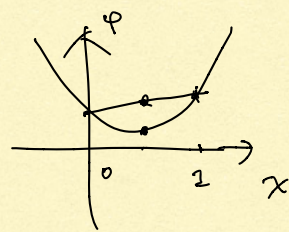


Last time :

(ref : Rick Durrett
Probability theory &
Examples)



• Jensen's inequality :

① if $\varphi: \mathbb{R} \rightarrow \mathbb{R}$ is a convex function.

X : is a random variable.

$$\mathbb{E}(\varphi(X)) \geq \varphi(\mathbb{E}X).$$

Hint : try the R.V. X , that values in $\{0,1\}$.

with equal probability $\frac{1}{2}$.

$$\mathbb{E}X = \frac{1}{2}. \quad \mathbb{E}(\varphi(X)) = \frac{1}{2} \cdot \varphi(0) + \frac{1}{2} \varphi(1)$$

• Hölder Inequality. (baby example : Cauchy Inequality).

• X, Y random variables :

$$X, Y : \Omega \rightarrow \mathbb{R}.$$

↑ sample space.

• let $1 \leq p < \infty$. $\|X\|_p = \{\mathbb{E}(|X|^p)\}^{\frac{1}{p}}$
p-norm

$$\left[\begin{array}{l} p=2, \quad \cdot \quad \| (x, y, z) \|_2 = (x^2 + y^2 + z^2)^{\frac{1}{2}} \quad \text{length.} \\ \cdot \quad \text{function } f \text{ on } [0,1] : \quad \|f\|_p = \left(\int_0^1 |f|^p dx \right)^{\frac{1}{p}} \end{array} \right]$$

• statement :

$$\mathbb{E} |X \cdot Y| \leq \|X\|_p \cdot \|Y\|_q.$$

$$p, q \text{ satisfies } \frac{1}{p} + \frac{1}{q} = 1.$$

$$\left[\text{Ex Cauchy ineq : } \mathbb{E} (|X \cdot Y|) \leq \|X\|_2 \cdot \|Y\|_2. \right]$$

$\Omega = \{1, 2, 3, \dots, n\}$, \mathbb{P} on Ω is equip-distribution.

$$X : \Omega \rightarrow \mathbb{R}. \quad X(i) = x_i$$

$$Y(i) = y_i.$$

$$\begin{aligned} \text{LHS} = \mathbb{E} (|X \cdot Y|) &= \sum_{i \in \Omega} \mathbb{P}(i) \cdot |X(i) Y(i)| \\ &= \sum_{i=1}^n \frac{1}{n} |x_i \cdot y_i|. \end{aligned}$$

$$\|X\|_2^2 = \mathbb{E}(|X|^2) = \sum_{i=1}^n \frac{1}{n} \cdot |x_i|^2;$$

"L²-norm"
of X

$$\|Y\|_2^2 = \sum \frac{1}{n} |y_i|^2.$$

LHS \leq RHS becomes.

$$\sum_i \frac{1}{n} |x_i y_i| \leq \left(\sum_i \frac{1}{n} |x_i|^2 \right)^{\frac{1}{2}} \cdot \left(\sum_i \frac{1}{n} |y_i|^2 \right)^{\frac{1}{2}}.$$

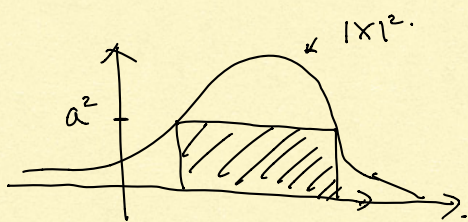
$$\Leftrightarrow \left(\frac{1}{n} \sum_i |x_i y_i| \right)^2 \leq \left(\frac{1}{n} \sum_i |x_i|^2 \right) \cdot \left(\frac{1}{n} \sum_i |y_i|^2 \right).$$

$$\boxed{\Leftrightarrow \left(\sum_i |x_i y_i| \right)^2 \leq \left(\sum_i |x_i|^2 \right) \left(\sum_i |y_i|^2 \right)}$$

• Markov Inequality: (Chebyshev inequality).

(1) $X: \Omega \rightarrow \mathbb{R}$, $a > 0$

$$a^2 \mathbb{P}(|X| > a) \leq \mathbb{E}(X^2).$$



$$\text{RHS} = \sum_{i \in \Omega} P_i \cdot (X(i))^2 \dots \textcircled{1}$$

$$\geq \sum_{i \in A} P_i (X(i))^2 \dots \textcircled{2}$$

$$A \subset \Omega, \quad A = \{i \in \Omega \mid |X(i)| \geq a\} \geq \sum_{i \in A} P_i \cdot a^2 \dots \textcircled{3}$$

$$= a^2 \cdot \mathbb{P}(A)$$

$$= \text{LHS}$$

(2). take any. $\varphi: \mathbb{R} \rightarrow \mathbb{R}$, $\varphi > 0$. (general statement).

$A \subset \mathbb{R}$ any subset.

$$\cdot \hat{v}_A = \inf_{\text{"min"}} \{ \varphi(y) : y \in A \}.$$

$$\cdot \frac{\hat{v}_A \cdot \mathbb{P}(X \in A)}{\textcircled{3}} \leq \frac{\mathbb{E}(\varphi(X) : X \in A)}{\textcircled{2}} \leq \frac{\mathbb{E}(\varphi(X))}{\textcircled{1}}$$