

Inequalities

Jensen's Inequality: Assume that φ is convex on (a, b) and that the random variable X takes values in (a, b) . Assume also that $X, \varphi(X)$ are integrable. Then

$$\varphi(E(X)) \leq E[\varphi(X)]$$

Lyapunov's inequality: Assume that X is a r.v.

(i) If $p \geq 1$, then

$$E[|X|]^p \leq E[|X|^p]$$

(ii) If $1 \leq p \leq q$, then

$$E[|X|^p]^{1/p} \leq E[|X|^q]^{1/q}$$

$$\|X\|_p \leq \|X\|_q$$

Proof: (i) Apply Jensen's inequality to X and $\varphi(x) = |x|^p$.

$$E[|X|]^p = \varphi(E(|X|)) \leq E[\varphi(|X|)] = E[|X|^p]$$

(ii) Apply Jensen's inequality to $Y = |X|^p$ and $\varphi(x) = |x|^{q/p}$.
convex since $q/p \geq 1$.

$$E[|X|^p]^{q/p} = E[Y]^{q/p} = \varphi(E[Y]) \leq E[\varphi(Y)] = E[(|X|^p)^{q/p}]$$

Hence

$$E[|X|^p]^{q/p} \leq E[|X|^q], \text{ take both sides to the power } 1/q$$

$$E[|X|^p]^{1/p} \leq E[|X|^q]^{1/q}$$

Modes of convergence

Question: What should it mean for a sequence $\{X_n\}$ of random variables to converge to a random variable X ?

↳ lot of different things.

Pointwise convergence: $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$ for all $\omega \in \Omega$.

Almost sure / a.s. convergence: There is a null set N such that $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$ for all $\omega \notin N$. $P(N) = 0$.

Convergence in probability: For $\epsilon > 0$,

$$P\{\omega : |X(\omega) - X_n(\omega)| \geq \epsilon\} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Convergence in mean / L^1 : $\lim_{n \rightarrow \infty} E[|X - X_n|] = 0$.

Convergence in L^p ($p \geq 1$): $\lim_{n \rightarrow \infty} E[|X - X_n|^p] = 0$.

Convergence in distribution: Let F_n and F be the distribution functions of X_n and X .

$$\lim_{n \rightarrow \infty} F_n(x) = F(x) \text{ at all points where } F \text{ is continuous.}$$

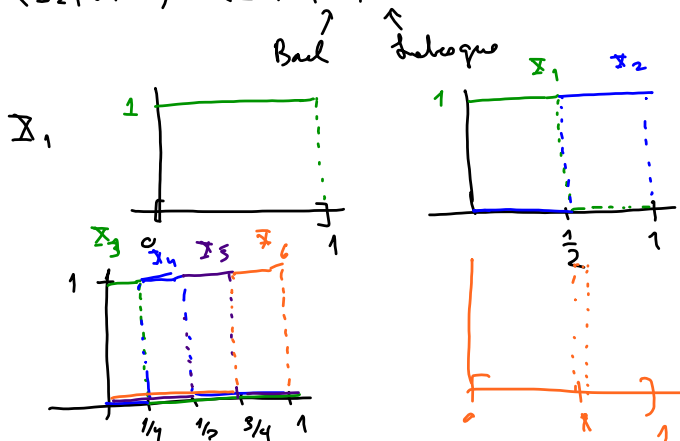
Advantage: The random variables can be defined on different prob. space.

Example: A sequence converging in prob and L^1 , but not converging pointwise.

$$X_n \rightarrow 0 \text{ in prob and } L^1$$

$$X_n \not\rightarrow 0 \text{ pointwise}$$

$$(\Omega, \mathcal{F}, P) = ([0, 1], \mathcal{B}, P)$$



Proposition: If $X_n \rightarrow X$ in L^p , then $X_n \rightarrow X$ in probability.

Proof: Chebyshev's inequality

$$P\{\omega: |X - X_n| \geq \varepsilon\} \leq \frac{1}{\varepsilon^p} E[|X - X_n|^p] \rightarrow 0,$$

hence convergence in probability.

Proposition: If $X_n \rightarrow X$ almost surely, then $X_n \rightarrow X$ in probability.

Proof: Note that there is a set Λ of prob. 0 such that $\lim_{n \rightarrow \infty} X_n(\omega) = X(\omega)$. Let $Z_n(\omega) = \sup\{|X_k(\omega) - X(\omega)|: k \geq n\}$

If $\omega \notin \Lambda$, then $\lim_{n \rightarrow \infty} Z_n(\omega) = 0$. Hence given an $\varepsilon > 0$,

let $T_n^\varepsilon = \{\omega: Z_n(\omega) \geq \varepsilon\}$. Then

$$\bigcap_{n \in \mathbb{N}} T_n^\varepsilon \subseteq \Lambda.$$

$$\text{hence } 0 = P\left(\bigcap_{n \in \mathbb{N}} T_n^\varepsilon\right) \stackrel{\text{cont. meas.}}{=} \lim_{n \rightarrow \infty} P(T_n^\varepsilon)$$

But then

$$P\{\omega: |X_n(\omega) - X(\omega)| \geq \varepsilon\} \stackrel{?}{\leq} P(T_n^\varepsilon) \rightarrow 0$$

Hence $X_n \rightarrow X$ in probability.

Problems

2.37, 3.2, 3.3, 3.4, 3.5, 3.6
 3 1 2.

3.3: Show that if $X \geq 0$ a.s., then $E[X] = 0$ iff $X = 0$ a.s.

Assume $X = 0$ a.s. Then $E[X] = E[0] = 0$.

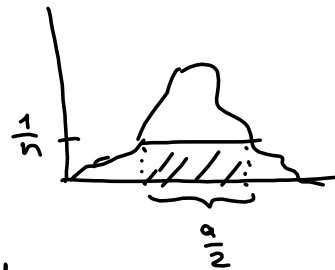
Assume X is not equal to 0 a.s. Then

$P\{\omega: X(\omega) > 0\} = a > 0$. Note

$$\{\omega: X(\omega) > 0\} = \bigcup_{n \in \mathbb{N}} \{\omega: X(\omega) > \frac{1}{n}\} \quad \leftarrow \text{increasing}$$

$a = P\{\omega: X(\omega) > 0\} = \lim_{n \rightarrow \infty} P\{\omega: X(\omega) > \frac{1}{n}\}$, so for large enough n , $P\{\omega: X(\omega) > \frac{1}{n}\} > \frac{a}{2}$.

$$E[X] \geq \frac{a}{2} \cdot \frac{1}{n}.$$



3.5 Show that if X is integrable, then

$$\lim_{x \rightarrow \infty} x P[|X| \geq x] = 0.$$

By replacing X by $|X|$ if necessary, we may assume that $X \geq 0$.

$$x P[X \geq x] \leq x P[\bar{X}_0 \geq x] = x \sum_{k \geq x} P[\bar{X}_0 = k]$$

$$= \sum_{k \geq x} x P[\bar{X}_0 = k] \leq \sum_{k \geq x} k P[\bar{X}_0 = k]$$

Take a
 convergent sequence
 $E[\bar{X}_0] = \sum_{k=0}^{\infty} k P[\bar{X}_0 = k] < \infty$
 since \bar{X}_0 integrable

$$\text{Hence } \lim_{x \rightarrow \infty} \sum_{k \geq x} k P[\bar{X}_0 = k] = 0.$$

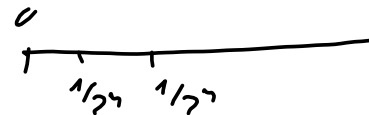
Sketch of 3.4: $X \geq 0$ r.v. F distribution function

$$\underline{E[X]} = \int_0^{\infty} (1-F(x)) dx$$

$$E[X] \leftarrow E[\bar{X}_n] = \sum_{k=0}^{\infty} \frac{k}{2^n} P\left(\frac{k-1}{2^n} < X \leq \frac{k}{2^n}\right)$$

$$= \frac{1}{2^n} \sum_{k=0}^{\infty} k \left\{ \underbrace{P\left[X > \frac{k-1}{2^n}\right]}_{l_k} - \underbrace{P\left[X > \frac{k}{2^n}\right]}_{l_{k+1}} \right\} \quad a_k = 1$$

$$= \frac{1}{2^n} \sum_{k=0}^{\infty} \underbrace{1}_{a_k} \cdot \underbrace{P\left[X > \frac{k-1}{2^n}\right]}_{l_k}$$



$$= \frac{1}{2^n} \sum_{k=1}^{\infty} \left(1 - P\left[X \leq \frac{k-1}{2^n}\right]\right) = \sum_{k=1}^{\infty} \underbrace{\left(1 - F\left(\frac{k-1}{2^n}\right)\right)}_{\text{Riemann sum}} \frac{1}{2^n}$$

$$\rightarrow \int_0^{\infty} (1-F(u)) du$$