

CONSEQUENCES OF CONVERGENCE IN DISTⁿ

Theorem: (Helly-Bray) : Let $(\Omega, \mathcal{A}, P)$ be a probability space
 Let $X_n \xrightarrow{d} X$ (ie $F_n \rightarrow F$)

1) If g is bdd and continuous a.s F (w.r.t a meas $\mu \equiv F$) then

$$E[g(X_n)] = \int g dF_n \xrightarrow{n \rightarrow \infty} E[g(X)] = \int g dF$$

[However note $X_n \xrightarrow{d} X \not\Rightarrow E(X_n) \rightarrow E(X)$ always]

2) If $E[g(X_n)] \xrightarrow{n \rightarrow \infty} E[g(X)]$ for any bdd and continuous g ,

then $X_n \xrightarrow{d} X$ ie $F_n \xrightarrow{n \rightarrow \infty} F$

Theorem: (Skorokhod) If $X_n \xrightarrow{d} X$ then $\exists \{Y_n\}_n$ and Y defined on some $(\Omega, \mathcal{A}, P)$ such that

a) $X_n \cong Y_n$ (equal in distⁿ)

b) $X \cong Y$

c) $Y_n \xrightarrow[P]{a.s.} Y$ (a.s. in prob)

{ eg: $X_1, X_2, \dots, X_n \sim \text{iid } f(x)$ }

Proof of Helly-Bray: Since $x_n \xrightarrow{d} x$ we use &

$$E[g(x_n)] = \int g d\mu_n = \int g dF_n$$

Also since $x_n \cong y_n$ we have $E[g(x_n)] = E[g(y_n)]$

Since $x \cong y$ we have $E[g(x)] = E[g(y)]$

We need to show $\int g(y_n) dP \longrightarrow \int g(y) dP$

We use DCT and so we need to show ① $|g(y_n)| \leq M$

$$\textcircled{2} \int |M| dP < \infty$$

We have g is bdd & continuous a.s $F \Rightarrow |g(y_n)| \leq M$
for some M

$$\text{Also } \int |M| dP = |M| \int dP = |M| < \infty$$

We have g is continuous a.s P_x

We also need to show $g(y_n) \xrightarrow{\text{a.s.}} g(y)$

ie we need to show $g(y_n(\omega)) \longrightarrow g(y(\omega)) \quad \forall \omega \in A \text{ st } P(A)=1$

[$y_n(\omega) = y_n$ some seq of # & we have $y_n \rightarrow y$ we need to show]
 $g(y_n) \rightarrow g(y)$

$$\text{let } A_1 \equiv \{\omega \in \Omega : y_n(\omega) \longrightarrow y(\omega)\}$$

$$\therefore P(A_1) = 1 \quad \text{since } y_n \xrightarrow{\text{a.s.}} y$$

$$\text{let } A_2 \equiv \{\omega \in \Omega \mid g \text{ is cont at } y(\omega)\}$$

$$\begin{aligned}
 P(A_2) &= P_Y(\{y \in \mathbb{R} : g \text{ is continuous at } y\}) \\
 &= P_X(\{y \in \mathbb{R} : g \text{ is cont at } y\}) \\
 &= 1.
 \end{aligned}$$

Take $A = A_1 \cap A_2$

$$0 \leq P(A^c) = P(A_1^c \cup A_2^c) \leq P(A_1^c) + P(A_2^c)$$

$$0 \leq P(A^c) \leq 0$$

$$\Rightarrow P(A^c) = 0 \quad \& \quad \text{so} \quad P(A) = 1$$

by continuity $y_n \xrightarrow{a.s.} y \Rightarrow g(y_n) \xrightarrow{a.s.} g(y)$

Mann Wald theorem: If $X_n \xrightarrow{d} X$ and g is continuous a.s F
 then $g(X_n) \xrightarrow{d} g(X)$
 eg: $X_n \xrightarrow{d} N(0,1)$ then $X_n^2 \xrightarrow{d} [N(0,1)]^2 = X_1^2$

Proof: $X_n \xrightarrow{d} X$ so $\exists \{y_n\} \& y$ st

$$X_n \approx y_n$$

$$X \approx y$$

$$y_n \xrightarrow{a.s.} y$$

$$\therefore g(y_n) \xrightarrow{a.s.} g(y)$$

$$\Rightarrow g(y_n) \xrightarrow{d} g(y)$$

$$\Rightarrow g(X_n) \xrightarrow{d} g(X)$$

{by cont of g

$$\{ \xrightarrow{a.s.} \Rightarrow \xrightarrow{d} \}$$

$$(\Omega, \mathcal{A}, P) \xrightarrow{Y} (\mathbb{R}, \mathcal{B})$$

P_Y is the induce meas & has 1-1 corresp with F_Y

By constructⁿ

$$F_Y = F_X$$

DEF: Convergence in L_2

X_n converges in r^{th} mean to X if $X_n \xrightarrow{L_1} X$ i.e. $E(|X_n - X|^r) \rightarrow 0$

Eg: for $r=2 \Rightarrow E(X_n^2) \xrightarrow{n \rightarrow \infty} E(X^2)$

for $r=1 \Rightarrow E(X_n) \rightarrow E(X)$

UNIFORM INTEGRATION

Def1: $\{X_t\}_{t \in T}$ is called integrable if $\sup_{t \in T} \{E|X_t|\} < \infty$

NOTE: It is possible to have $E|X_t| < \infty$ but $\sup_{t \in T} (E|X_t|) = \infty$

eg if $E(|X_t|) = t$

Def2: $\{X_t\}_{t \in T}$ is called uniformly integrable (u.i.) if

$$\sup_{t \in T} E|X_t| I_{[|X_t| \geq \lambda]} \xrightarrow{\lambda \rightarrow \infty} 0 \quad \left\{ \begin{array}{l} \text{i.e. } \forall \varepsilon > 0 \exists \lambda_\varepsilon \text{ st} \\ \sup_{t \in T} E|X_t| I_{[|X_t| \geq \lambda]} \leq \varepsilon \quad \forall \lambda \geq \lambda_\varepsilon \end{array} \right.$$

i.e. if for large values of X_t the $E|X_t|$ is also large then the function is not integrable.

Remark: Suppose that $\{X_t\}_{t \in T}$ is such that $|X_t| \leq Y$, $\forall t \in T$ and $Y \in L_1$ for some Y then $\{X_t\}$ is uniformly integrable.

Proof

$$E|X_t| I_{[|X_t| > \lambda]} = \int_{[|X_t| > \lambda]} |X_t| d\mu \leq \int_{[|X_t| > \lambda]} |Y| d\mu$$

$$E |x_t| I_{[|x_t| > \lambda]} \leq \int_{[|y| > \lambda]} |y| d\mu$$

We now need to show $\mu[|y| > \lambda] \rightarrow 0$ as $\lambda \rightarrow \infty$

Now $\mu[|y| > \lambda] \leq \frac{E |y|}{\lambda}$ by Markov inequality

$$\mu[|y| > \lambda] \leq \left(\frac{E |y|}{\lambda} \right) \rightarrow 0 \text{ as } \lambda \rightarrow \infty$$

$$\therefore \sup_t E |x_t| I_{[|x_t| > \lambda]} \rightarrow 0 \text{ as } \lambda \rightarrow \infty.$$

THEOREM Uniform Absolute Continuity of the Integral

Assume $\mu(\Omega) < \infty$ *

$\{x_t\}_{t \in T}$ is u.i $\iff$ (a) $\{x_t\}_{t \in T}$ is integrable

(b) $\mu(A) < \delta_\epsilon \implies \sup_t \int_A |x_t| d\mu < \epsilon$

Proof : " $\implies$ " a) $E |x_t| = E |x_t| I_{[|x_t| \leq \lambda]} + E |x_t| I_{[|x_t| > \lambda]}$

$$\leq \lambda \mu(\Omega) + E |x_t| I_{[|x_t| > \lambda]}$$

$$\leq \lambda \mu(\Omega) + \epsilon$$

$$< \infty$$

{ since $\mu(\Omega) < \infty$

$\Rightarrow$ b) Take A st $\mu(A) < \delta$ for δ to be chosen later.

Note $[|x_t| \geq \lambda]$ is one possible choice for A .

Consider
$$\int_A |x_t| d\mu = \int_A |x_t| I_{[|x_t| \geq \lambda]} + \int_A |x_t| I_{[|x_t| < \lambda]} d\mu.$$

$$\leq \int_A |x_t| I_{[|x_t| \geq \lambda]} + \lambda \mu(A)$$

$$\leq \int_{\Omega} |x_t| I_{[|x_t| \geq \lambda]} + \lambda \mu(A).$$

$$\sup_{t \in T} \int_A |x_t| d\mu \leq \sup_{t \in T} \int |x_t| I_{[|x_t| \geq \lambda]} d\mu + \lambda \mu(A)$$

$$< \frac{\varepsilon}{2} + \lambda \mu(A) \quad \left\{ \text{by def of u.i} \right.$$

$$< \frac{\varepsilon}{2} + \lambda \frac{\varepsilon}{2\lambda} \quad \left\{ \text{letting } \delta = \frac{\varepsilon}{2\lambda} \right.$$

$$\therefore \sup_{t \in T} \int_A |x_t| d\mu < \varepsilon$$

$\Leftarrow$ Take $A = [|x_t| \geq \lambda]$

$$\text{Consider } \mu(A) = \mu(|x_t| \geq \lambda) \leq \frac{E(|x_t|)}{\lambda}$$

Now by (a) $\{x_t\}$ is integrable and so $E|x_t| < \infty \Rightarrow \frac{E|x_t|}{\lambda} \xrightarrow{\lambda \rightarrow \infty} 0$

$\therefore \mu(A) \rightarrow 0$ and so by (b) we get $\sup_{t \in T} \int_A |x_t| d\mu < \varepsilon$

Theorem : (Vitali) Let $r > 0$, assume $\{x_n\}$ st $x_n \in L_r \equiv \{g: \int |g|^r < \infty\}$
 Assume $x_n \xrightarrow{\mu} x^*$ for some x and $\mu(L) < \infty^*$ then the foll
 statements are equivalent

- 1) $\{|x_n|^r, n \geq 1\}$ are u.i
- 2) $x_n \xrightarrow{L_r} x$
- 3) $\int |x_n|^r \longrightarrow \int |x|^r$
- 4) $\overline{\lim} \int |x_n|^r \leq \int |x|^r < \infty$

Corollary (L_1 convergence)

$$\underline{x_n \xrightarrow{L_1} x} \iff \begin{array}{l} \text{a) } x_n \xrightarrow{\mu} x \quad ; \quad \mu(L) < \infty \\ \text{b) } \{|x_n|, n \geq 1\} \text{ are u.i} \end{array}$$

Proof " $\Leftarrow$ " $x_n \xrightarrow{\mu} x$ and $\{|x_n|\}$ are u.i

so by Vitali's theorem with $r=1$ we get $x_n \xrightarrow{L_1} x$

" $\Rightarrow$ " a) $\mu(|x_n - x| > \varepsilon) \leq \frac{\int |x_n - x|}{\varepsilon} \longrightarrow 0$
 since $x_n \xrightarrow{L_1} x$
 $\Rightarrow x_n \xrightarrow{\mu} x$

$\therefore$ by Vitali's theorem $\Rightarrow \{|x_n|\}$ are u.i

Proof of Vitali's Theorem :

Plan: Show $1) \Rightarrow 2) \Rightarrow 3) \Rightarrow 4) \Rightarrow 1)$

$2) \Rightarrow 3)$ trivial and done before.

$3) \Rightarrow 4)$ easy.

$4) \Rightarrow 1)$ Hard (read but no proof for exam)

Prove $1) \Rightarrow 2)$

To show X also $\in L^p$

We have $X_n \xrightarrow{M} X \Rightarrow \exists n'$ st $X_{n'} \xrightarrow{a.e} X$ (by theorem)

$$\Rightarrow |X_{n'}|^p \xrightarrow{a.s.} |X|^p$$

$$\text{Now } E(|X|^p) = E(\lim |X_{n'}|^p) \quad (\text{exercise})$$

$$\leq \lim E |X_{n'}|^p \quad (\text{Fatou's lemma})$$

$$< \infty \quad (\because \{|X_{n'}|^p\} \text{ are u.i.})$$

Now if $\{|X_n|^p\}$ is u.i. and $X \in L^p$ then $\{|X_n - X|^p\}$ is also u.i. (exercise)

To show now $E(|X_n - X|^p) \rightarrow 0$

$$E |X_n - X|^p = E |X_n - X|^p I_{[|X_n - X| \geq \varepsilon]} + E |X_n - X|^p I_{[|X_n - X| \leq \varepsilon]}$$

$$\leq E |X_n - X|^p I_{[|X_n - X| \geq \varepsilon]} + \varepsilon^p$$

$$< \varepsilon + \varepsilon^p \quad (\text{exercise})$$

To show uniform integrability we need to show

$$\sup E |X_n| I_{[|X_n| > \lambda]} \rightarrow 0 \quad \text{as } \lambda \rightarrow \infty$$

$\Leftrightarrow$

② $\sup_n E |X_n| < \infty$

③ $\forall \varepsilon > 0, \exists \delta > 0$ st $\sup_n \int_A |X_n| d\mu < \varepsilon$ whenever $\mu(A) < \delta$

But Vitali's Theorem gives simpler conditions.

Proof

To show (1) $\Rightarrow$ (2)

(2) $E |X_n - X|^r \rightarrow 0$ as $n \rightarrow \infty$

ie given $\varepsilon > 0 \exists M_\varepsilon$ st $E |X_n - X|^r < \varepsilon \quad \forall n \geq M_\varepsilon$

Let $\varepsilon' > 0$ and we will define it later

$$\textcircled{C} = E |X_n - X|^r = \underbrace{E |X_n - X|^r I_{[|X_n - X| > \varepsilon']}}_{\textcircled{A}} + \underbrace{E |X_n - X|^r I_{[|X_n - X| \leq \varepsilon']}}_{\textcircled{B}}$$

Clearly $E |X_n - X|^r I_{[|X_n - X| \leq \varepsilon']} \leq (\varepsilon')^r$

Now we have $X_n \xrightarrow{\mu} X \Rightarrow \exists n'$ st $X_{n'} \xrightarrow{a.s.} X$

ie $\lim_{n \rightarrow \infty} X_{n'} = X = \underline{\lim} X_{n'} = \overline{\lim} X_n$

Also if $\lim_{n \rightarrow \infty} X_{n'} = X$ then $\lim_{n \rightarrow \infty} |X_{n'}|^r = |X|^r$

$\therefore \lim |X_{n'}|^r = |X|^r$

Now $E |X|^r = E \lim |X_{n'}|^r \leq \lim E |X_{n'}|^r \leq \sup_n E |X_n|^r < \infty$

Fatou's lemma

by ①

?

$$|x_n|^n \text{ are u.i.} \Rightarrow |x_n - x|^n \text{ is u.i.} \quad (\text{claim 6})$$

$$\forall x \in L_2$$

Since $x_n \xrightarrow{P} x$, given any $\varepsilon, \delta \exists N(\delta, \varepsilon)$ st $P(|x_n - x| > \varepsilon) < \delta \quad \forall n \geq N(\varepsilon, \delta)$

Now since $\{|x_n - x|^n\}_{n \geq 1}$ is u.i. given any $\varepsilon \exists \delta$ st $\mu(A) < \delta$

$$\text{then } \sup_n \int_A |x_n - x|^n dP < \varepsilon \quad \left\{ \text{by def of u.i.} \right.$$

$$\therefore \text{ for } \varepsilon' \exists \delta' \text{ st } \mu(A) < \delta' \text{ and } \sup_n \int_A |x_n - x|^n dP < \varepsilon'$$

we also have ε', δ' st $\exists N(\delta', \varepsilon')$ and $P(|x_n - x| > \varepsilon') < \delta' \quad \forall n \geq N(\delta', \varepsilon')$

Now take a fixed $n^* \geq N(\varepsilon', \delta')$, we have $P(|x_{n^*} - x| > \varepsilon') < \delta'$

$$\Rightarrow \sup_n \int_{\{|x_{n^*} - x| > \varepsilon'\}} |x_n - x|^n dP < \varepsilon' \quad \left[\text{let } A = \{|x_{n^*} - x| > \varepsilon'\} \right]$$

$$\text{Also } \int_{\{|x_{n^*} - x| > \varepsilon'\}} |x_{n^*} - x|^n dP \leq \sup_n \int_{\{|x_{n^*} - x| > \varepsilon'\}} |x_n - x|^n dP < \varepsilon'$$

$$\text{ie } \int_{\{|x_n - x| > \varepsilon'\}} |x_n - x|^n dP < \varepsilon' \quad \forall n \geq N(\varepsilon', \delta')$$

$$\text{ie } E |x_n - x|^n I_{[|x_n - x| > \varepsilon']} < \varepsilon'$$

$$\therefore \textcircled{C} < \varepsilon' + (\varepsilon')^n \Rightarrow \textcircled{C} < \varepsilon \quad \left[\text{let } \varepsilon = \varepsilon' + (\varepsilon')^n \right]$$

★

THEOREM

let $S_\varepsilon = \left\{ \sum y_j I_{I_j} \mid I_j \text{ are disjoint finite intervals} \right\}$

Then $\forall \varepsilon > 0, \exists \gamma_\varepsilon \in S_\varepsilon$ st $\int |x - \gamma_\varepsilon| d\mu_F < \infty$ for any $x \in L_1$

NOTE: Any functⁿ x in L_1 can be approximated by a piecewise functⁿ

eg: let x be a density functⁿ which is in L_1 (wrt Lebesgue meas λ)
then you can use histograms to approximate this density.

Proof let $x \geq 0$.

Recall Prop 2.2.3: $\forall \varepsilon > 0 \exists Z_\varepsilon$ a simple functⁿ st

$$|x(\omega) - Z_\varepsilon(\omega)| \leq \varepsilon \quad \forall \omega \in \Omega.$$

$$Z_\varepsilon = \sum_{i=1}^n z_i I_{A_i} \quad \text{st } z_i \in \mathbb{R} \text{ \& } \sum_{i=1}^n A_i = \mathcal{B}$$

Note: The diff b/w Z_ε & γ_ε is that here A_i are disjoint borel sets and not disjoint finite intervals

Now μ_F is a L-S measure.

Note $\mu(A_i) < \infty \quad \forall i=1, \dots, n$ for x in L_1

$\downarrow$ Proof Assume $\exists i_0$ st $\mu(A_{i_0}) = \infty$.

$$\text{Consider } \int |x| d\mu \leq \int |x - Z_\varepsilon| d\mu + \int |Z_\varepsilon| d\mu.$$

$$= \int |x - Z_\varepsilon| d\mu + \int \left| \sum_{i=1}^n z_i I_{A_i} \right| d\mu$$

$$\leq \int |x - Z_\varepsilon| d\mu + \int \sum |z_i| I_{A_i} d\mu$$

$$= \int |x - Z_\varepsilon| d\mu + \sum z_i \mu(A_i)$$

$$= \infty$$

which is a $\rightarrow \leftarrow$ since $x \in L_1$ $\nwarrow$

Note also that $\int |x - z_\varepsilon| d\mu < \infty$

↙ since
$$\int |x - z_\varepsilon| d\mu \leq \int |x| + \int |z_\varepsilon| < \infty$$

↗ since $x \in L_1$
& $\mu(A_i) < \infty \quad \forall A_i$

Choose z_ε st $\int |x - z_\varepsilon| < \varepsilon/2$

↙ (*)

↗ since $\int |x - z_\varepsilon| \leq M$
we can rescale so it is $\leq \varepsilon/2$

Pg 16 Ex 1.2.3 Halmos approximation lemma

↙ Here we have $\mathcal{B} = \sigma[\mathcal{C}_F]$

$\mathcal{C}_F \supset \left\{ \sum_n I_n : \text{each } I_n \text{ is intv of type } (a, b] \right\}$

Also $A_1, A_2, \dots, A_n \in \mathcal{B}$ and $\mu(A_i) < \infty \quad \forall A_i$ then

$\exists B_1, B_2, \dots, B_n \in \mathcal{C}_F$ st $\mu(A_i \Delta B_i) < \frac{\varepsilon}{2n|z_i|}$

where $A_i \Delta B_i = (A_i^c \cap B_i) \cup (B_i^c \cap A_i)$

Now we define $\gamma_\varepsilon = \sum_{i=1}^n z_i I_{B_i}$

Consider
$$\begin{aligned} \int |z_\varepsilon - \gamma_\varepsilon| d\mu &= \int \left| \sum_{i=1}^n z_i (I_{B_i} - I_{A_i}) \right| d\mu \\ &\leq \int \sum_{i=1}^n |z_i| |I_{B_i} - I_{A_i}| d\mu \\ &= \sum |z_i| \int (I_{B_i} - I_{A_i}) d\mu \end{aligned}$$

Now $|I_{B_i} - I_{A_i}| = \begin{cases} 1 & \text{if } \omega \in A_i \setminus B_i = A_i \cap B_i^c \\ 1 & \text{if } \omega \in B_i \setminus A_i = A_i^c \cap B_i \\ 0 & \text{if } \omega \in B_i \cap A_i \end{cases}$

$$\int |I_{B_i} - I_{A_i}| d\mu = \mu \left(\int 1 d\mu \right)_{A_i B_i^c \cup A_i^c B_i} = \mu(A_i B_i^c \cup A_i^c B_i).$$

$$\therefore \int |z_\varepsilon - y_\varepsilon| d\mu \leq \sum |z_i| \mu(A_i \Delta B_i)$$

$$\leq \sum_{i=1}^n |z_i| \frac{\varepsilon}{2n|z_i|}$$

$$= \sum_{i=1}^n \frac{\varepsilon}{2n}$$

$$= \frac{\varepsilon}{2}$$

— (**)

$$\text{Now } \int |x - y_\varepsilon| d\mu \leq \int |x - z_\varepsilon| + \int |z_\varepsilon - y_\varepsilon|$$

$$< \frac{\varepsilon}{2} + \frac{\varepsilon}{2}$$

from (*) & (**)

$$\text{Rewrite } y_\varepsilon = \sum_{i=1}^n z_i I_{B_i} = \underset{\text{rewrite}}{\sum_{j=1}^m y_j I_{I_j}}$$

where I_j are disjoint intervals

$$\text{eg: If } y_\varepsilon = 3 I_{[0,1] \cup (1,7)} + 2 I_{[0,1] \cup (1,8)}$$

$$= 5 I_{[0,1]} + 5 I_{(1,7)} + 2 I_{(7,8)}$$

5.7 → Theorem 5.7
know def & statements of all modes
in this theorem.

Inequalities (Statements)

Pg 51, 52, 53, 54, 55, 56

Proof on pg 58

Theorem 5.8 (Proof in class)
Whole statement

(No) theorem 5.6

Inequalities — Minkowski's inequality
(No proofs) Chebyshev's
Markov's
Other ones which were used in class.

Last 2 HW - Proof

Statements & Definition (10 pts)

Sanyay's Problem