

Lecture 22: Martingale convergence

(1)

Under mild conditions subMs converge a.e.

Dobbs upcrossing inequality is the key to the proof.

Here is a motivation in terms of how to

check a sequence of numbers x_1, x_2, x_3, \dots converges.

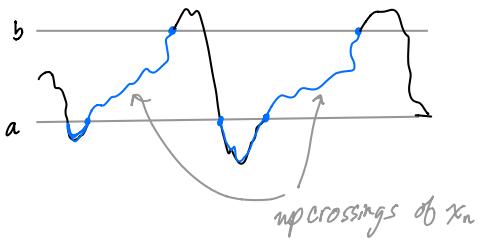
x_n does not converge

$$\text{iff } \liminf_n x_n < \limsup_n x_n$$

$$\text{iff } \exists a, b \in \mathbb{Q} \text{ s.t. } a < b$$

$x_n > b$ & $x_n < a$ infinitely often.

i.e.



iff $\exists a, b \in \mathbb{Q}$ s.t. x_n has infinitely many upcrossings of $[a, b]$

Dobbs upcrossings

Let (X_1, X_2, \dots, X_n) be adapted to the filtration $(\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n)$.

Fix $-\infty < a < b < \infty$ and define

$$\alpha_1 = \min \left(\{k \geq 1 : X_k \leq a\} \cup \{n\} \right) \quad \left\{ \begin{array}{l} \text{upcrossings} \\ \text{of } [a, b] \end{array} \right.$$

$$\beta_1 = \min \left(\{k > \alpha_1 : X_k \geq b\} \cup \{n\} \right)$$

$$\alpha_2 = \min \left(\{k > \beta_1 : X_k \leq a\} \cup \{n\} \right) \quad \left\{ \begin{array}{l} \text{upcrossings} \\ \text{of } [a, b] \end{array} \right.$$

$$\beta_2 = \min \left(\{k > \alpha_2 : X_k \geq b\} \cup \{n\} \right) \quad \left\{ \begin{array}{l} \text{upcrossings} \\ \text{of } [a, b] \end{array} \right.$$

:

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The number of upcrossings of $[a, b]$, denoted $U_{a,b}$, is defined as

$$U_{a,b} := \sum_{j=1}^n \mathbb{I}_{\{X_{\alpha_j} \leq a, X_{\beta_j} \geq b\}} \quad \leftarrow \begin{array}{l} \text{can't be more than} \\ n \text{ terms} \end{array}$$

Proposition:

$\alpha_1, \dots, \alpha_n, \beta_1, \dots, \beta_n$ are stopping times w.r.t $(\mathcal{F}_1, \dots, \mathcal{F}_n)$ and U_{ab} is \mathcal{F}_n -measurable

Proof:

We've already shown that α_i is a ST.

To show β_i is a stopping time notice that

- If $1 \leq m < n$ then

$$\begin{aligned} \{\beta_1 = m\} &= \{\alpha_1 < m, \beta_1 = m\} \\ &= \bigcup_{j=1}^{m-1} \{\alpha_1 = j, \beta_1 = m\} \\ &= \bigcup_{j=1}^{m-1} \{\alpha_1 = j, X_{j+1} < b, \dots, X_{m-1} < b, X_m \geq b\} \\ &\quad \uparrow \qquad \qquad \qquad \uparrow \qquad \qquad \qquad \uparrow \\ &\in \mathcal{F}_j \subset \mathcal{F}_m \qquad \qquad \qquad \in \mathcal{F}_m \\ &\in \mathcal{F}_m \end{aligned}$$

- If $m = n$ then

$$\{\beta_1 = n\} = \{\beta_1 < n\}^c = \left(\bigcup_{j=1}^{n-1} \{\beta_1 = j\} \right)^c \in \mathcal{F}_n$$

An induction argument shows $\alpha_2, \beta_2, \dots, \alpha_n, \beta_n$ are ST.

$\therefore X_{\alpha_j}$ is \mathcal{F}_{α_j} -measurable &

X_{β_j} is \mathcal{F}_{β_j} -measurable

Since \mathcal{F}_{α_j} & \mathcal{F}_{β_j} are sub σ -fields of \mathcal{F}_n

U_{ab} is \mathcal{F}_n -measurable.

QED.

Theorem: (Doob's upcrossing ineq)

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Suppose (X_1, X_2, \dots, X_n) is a non-negative subM w.r.t filtration $(\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n)$.

Let $c > 0$ & $U_{0,c}$ denote the upcrossings of $[0, c]$. Then

$$E(U_{0,c}) \leq \frac{E(X_n) - E(X_1)}{c}.$$

Proof: The idea is to write $X_n - X_1$ as a telescoping sum: must be X_n

$$\begin{aligned}
 X_n - X_1 &= (X_{\beta_n} - X_{\alpha_n}) + (X_{\alpha_{n-1}} - X_{\beta_{n-1}}) \\
 &\quad + (X_{\beta_{n-1}} - X_{\alpha_{n-1}}) + (X_{\alpha_{n-1}} - X_{\beta_{n-2}}) \\
 &\quad + \vdots \\
 &\quad + (X_{\beta_2} - X_{\alpha_2}) + (X_{\alpha_2} - X_{\beta_1}) \\
 &\quad + (X_{\beta_1} - X_{\alpha_1}) + (X_{\alpha_1} - X_1)
 \end{aligned}$$

$\beta_0 = 1$

Notice that

$$\begin{aligned}
 E(\text{a red term}) &= E(X_{\alpha_j} - X_{\beta_{j-1}}) \\
 &= E(E(X_{\alpha_j} - X_{\beta_{j-1}} | \mathcal{F}_{\beta_{j-1}})) \\
 &= E(E(X_{\alpha_j} | \mathcal{F}_{\beta_{j-1}}) - X_{\beta_{j-1}}) \\
 &\stackrel{\text{def}}{\geq} X_{\beta_{j-1}} \quad \text{by The} \\
 &\quad \text{Finite optional} \\
 &\quad \text{Sampling Thm} \\
 &\geq 0
 \end{aligned}$$

Moreover

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$$(X_{\beta_j} - X_{\alpha_j}) \begin{cases} > c & \text{if } (X_{\alpha_j}, X_{\beta_j}) \text{ an upper boundary} \\ 0 & \text{if } \beta_j = \alpha_j = n \\ > 0 & \text{if } \beta_j = n \text{ but } \alpha_j < n \\ & \text{since } X_{\alpha_j} = 0 \text{ & } X_{\beta_j} > 0 \text{ by non-negativity} \end{cases}$$

\therefore The sum of all blue terms $\geq c \mathcal{U}_{0,c}$

$$\therefore E(X_n - X_1) \geq E(\underbrace{\text{blue}}_{\geq U_{0,c}} + E(\underbrace{\text{red}}_{\geq 0}) \geq c E(U_{0,c}).$$

QED

Corollary:

If (X_1, \dots, X_n) is a subM m.r.b. $(\Theta_1, \dots, \Theta_n)$
 and $-\infty < a < b < \infty$ then

$$E(u_{a,b}) = \frac{E(X_n-a)^+ - E(X_1-a)^+}{b-a} \leq \frac{E(X_n^+) + a^-}{b-a}.$$

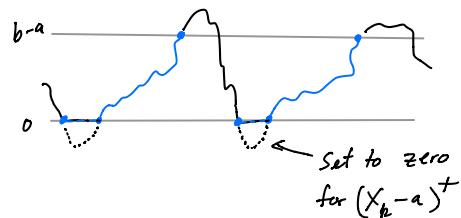
Proof:

For the first inequality:

$\#$ crossings of $[a, b]$ in (X_1, \dots, X_n)

upcrossings of $[a, b-a]$ in $(X_1 - a, \dots, X_n - a)$

upcrossings of $[0, b-a]$ in $((X_1 - a)^+, \dots, (X_n - a)^+)$



By Doob's result we therefore have

$$E(U_{a,b-a}) \leq \frac{E(X_{n-a})^+ - E(X_{i-a})^+}{b-a}$$

For the second inequality just notice (5)

$$\begin{aligned}
 E(X_{n-a})^+ - E(X_{-a})^+ &\leq E(X_n - a)^+ \\
 &= E(X_n + (-a))^+ \\
 &\leq E(X_n^+ + (-a)^+) \\
 &\quad \text{since } (-\cdot)^+ \\
 &\quad \text{is convex} \\
 &= E(X_n^+) + a^- \\
 &\quad \text{QED}
 \end{aligned}$$

a.e. Convergence of martingales

For this section let

$$\mathcal{F}_\infty := \sigma(\mathcal{F}_1, \mathcal{F}_2, \dots)$$

for a filtration $(\mathcal{F}_n)_{n \geq 1}$.

Theorem: $(E(X_n^+) \text{ bdd} \Rightarrow \text{a.e. conv})$

Let $(X_n)_{n \geq 1}$ be a subM w.r.t. filtration $(\mathcal{F}_n)_{n \geq 1}$.

If $\sup_n E(X_n^+) < \infty$ then $\exists X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$

s.t.

$$X_n \xrightarrow{P\text{-a.e.}} X_\infty$$

Proof:

For $-\infty < a < b < \infty$ let $U_{a,b}^n$ denote the number of upcrossings of $[a, b]$ from (X_1, \dots, X_n) , with $U_{a,b}^\infty$ for $(X_n)_{n \geq 1}$.

We must have that

$$0 \leq U_{a,b}^n \uparrow U_{a,b}^\infty \text{ as } n \rightarrow \infty$$

$$\begin{aligned}
 \therefore E(U_{a,b}^\infty) &= \lim_n E(U_{a,b}^n) \\
 &\leq \lim_n \frac{E(X_n^+) + a^-}{b-a} \\
 &= \frac{\sup_n E(X_n^+) + a^-}{b-a} < \infty
 \end{aligned}$$

$$\therefore U_{a,b}^\infty < \infty \text{ P-a.e.}$$

Now

$$\begin{aligned}
 P(\liminf_n X_n < \limsup_n X_n) \\
 &\leq P\left(\bigcup_{\substack{a < b \\ a, b \in \mathbb{R}}} \{U_{a,b}^\infty = \infty\}\right) \\
 &\quad \text{must } \exists ab \text{ with an infinite # of upcrossings.} \\
 &= 0 \text{ since } U_{a,b}^\infty < \infty \text{ P-a.e.}
 \end{aligned}$$

$$\therefore X_n \xrightarrow{P\text{-a.e.}} \limsup_n X_n =: X_\infty$$

must be $\in \mathcal{F}_\infty$ by closure Thm.

To see why $X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$

$$\begin{aligned}
 E|X_\infty| &= E|\liminf_n X_n| = E\left(\liminf_n |X_n|\right) \\
 &\stackrel{\text{Fatou}}{\leq} \liminf_n E|X_n| \\
 &\leq \sup_n E|X_n| \\
 &= \sup_n E(2X_n^+ - X_n) \\
 &\stackrel{\text{since } E(X_i) \leq E(X_n)}{\leq} \sup_n (2E(X_n^+) - \underbrace{E(X_i)}_{< \infty}) \\
 &\leq \sup_n (2E(X_n^+) - E(X_i)) \quad \text{by subM} \\
 &< \infty
 \end{aligned}$$

QED

Remark: Note the analog to monotonic sequences where $\sup_n E(X_n^+) < \infty$ plays the role of $\sup_n X_n^+ < \infty$

Another bdd type condition is that the X_n^+ 's are UI.

Proposition: $((X_n^+)_n \text{ UI} \Rightarrow E(X_n^+) \text{ bdd})$

If $X_1, X_2, \dots \in L_1(\Omega, \mathcal{F}, P)$ and $(X_n^+)_n$ are UI then $\sup_n E(X_n^+) < \infty$.

Proof:

$$\sup_n E(X_n^+) \leq \underbrace{\sup_n E(X_n^+ I_{X_n^+ \geq c})}_{\xrightarrow{c \rightarrow \infty} 0 \text{ by defn of UI}} + \underbrace{\sup_n E(X_n^+ I_{X_n^+ < c})}_{< c}$$

so for a large enough c this is $< \infty$.

QED.

Here is an application of the above proposition which comes in handy for subsequent results.

Theorem: (Lévy's Smoothing Martingale)

If $X \in L_1(\Omega, \mathcal{F}, P)$ & $(\mathcal{F}_n)_{n \geq 1}$ is a filtration then

$$E(X|\mathcal{F}_n) \xrightarrow{n \rightarrow \infty} E(X|\mathcal{F}_\infty) \text{ a.e. & in } L_1$$

where $\mathcal{F}_\infty = \sigma(\mathcal{F}_1, \mathcal{F}_2, \dots)$.

Proof:

$$\text{Let } X_n := E(X|\mathcal{F}_n).$$

We first show the X_n 's are UI.

Since $|X_n| = |E(X|\mathcal{F}_n)| \leq E(|X| |\mathcal{F}_n)$ it will be sufficient to show the $E(|X| |\mathcal{F}_n)$'s are UI.

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i.e. show

$$\lim_{c \rightarrow \infty} \sup_n \int E(|X| |\mathcal{F}_n) dP = 0$$

$E(|X| |\mathcal{F}_n) \geq c$

" "

$$\int |X| dP \quad \text{since } \{E(|X| |\mathcal{F}_n) \geq c\} \text{ is a } \mathcal{F}_n\text{-set}$$

$$\tilde{P}(E(|X| |\mathcal{F}_n) \geq c) \cdot \int_\Omega |X| dP$$

" " finite

where $\tilde{P}(\cdot) = \frac{\int_\Omega |X| dP}{\int_\Omega |X| dP}$ is a prob measure.

By Markov's meg

$$\tilde{P}(E(|X| |\mathcal{F}_n) \geq c) \leq \frac{E\{E(|X| |\mathcal{F}_n)\}}{c} = \frac{E|X|}{c} \xrightarrow{c \rightarrow \infty} 0$$

\therefore the X_n 's are indeed UI.

\therefore the X_n^+ 's are UI so the $E(X_n^+)$'s bdd

$\therefore \exists X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$ s.t.

$$X_n \xrightarrow{a.e.} X_\infty \text{ by subM a.e. Thm}$$

$\therefore X_n \xrightarrow{L_1} X_\infty$ by the L_p convergence Thm since the X_n 's are UI

To finish we show $X_\infty \stackrel{a.e.}{=} E(X|\mathcal{F}_\infty)$.

In particular show:

(i) X_∞ is \mathcal{F}_∞ -measurable ✓

(ii) $X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$ ✓

(iii) Integrates like X over \mathcal{F}_∞ -sets.

i.e. $E(X \mathbf{1}_A) = E(X_\infty \mathbf{1}_A) \forall A \in \mathcal{F}_\infty$

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For (iii) notice that $\forall A \in \bigcup_{n=1}^{\infty} \mathcal{F}_n$ (9)

$$E(XI_A) = E(X_n I_A) \quad \text{if large } n \\ \text{since } X_n = E(X|\mathcal{F}_n)$$

$\therefore \forall A \in \bigcup_{n=1}^{\infty} \mathcal{F}_n \leftarrow \text{a field generating } \mathcal{F}_{\infty}$

$$E(X_{\infty} I_A) = E\left(\lim_n X_n I_A\right) \quad \text{since } X_n \xrightarrow{a.s.} X_{\infty} \\ = \lim_n E(X_n I_A) \quad \text{by UI cond.} \\ \text{for } \lim = \lim \\ \stackrel{(*)}{=} E(X I_A) \quad \text{by (*)}$$

To show $E(X_{\infty} I_A) = E(X I_A) \quad \forall A \in \mathcal{F}_{\infty}$

let

$$\mathcal{Y} = \{A \in \mathcal{F}: E(X_{\infty} I_A) = E(X I_A)\}$$

By (*) we have

$$\text{field}^2 \rightsquigarrow \bigcup_{n=1}^{\infty} \mathcal{F}_n \subset \mathcal{Y} \quad \text{a } \mathcal{X}\text{-system}$$

$$\therefore \lambda \left\langle \bigcup_{n=1}^{\infty} \mathcal{F}_n \right\rangle \subset \mathcal{Y}$$

// Dynkin's $\mathcal{T}-\lambda$

$$\sigma \left\langle \bigcup_{n=1}^{\infty} \mathcal{F}_n \right\rangle$$

$$\therefore E(X_{\infty} I_A) = E(X I_A) \quad \forall A \in \mathcal{F}_{\infty}$$

by "Good sets".

QED

(10)

Another way to quantify "badness" of subMs is with "closers".

Definition: A pair (X_0, \mathcal{F}_0) ^{r.v.} ^{sub σ-field of F} closes

a subM $(X_n)_{n \geq 1}$ on the right if

$$X_1, X_2, \dots, X_n, \dots X_0$$

is a subM w.r.t. $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_n, \dots, \mathcal{F}_0$, i.e. if

$$E(X_0 | \mathcal{F}_n) \stackrel{P.a.e.}{\geq} X_n \quad \forall n \in \mathbb{N}.$$

Definition: (X_0, \mathcal{F}_0) is a nearest closer of $(X_n)_{n \geq 1}$ if

$$X_1, X_2, \dots, X_0, X_0$$

is a subM w.r.t. $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_0, \mathcal{F}_0$

for every closer (X_0, \mathcal{F}_0) of $(X_n)_{n \geq 1}$.

The existence of a closer is equivalent to the UI condition:

Theorem: $(\exists \text{closer} \iff X_n^+ \text{'s are UI})$

If $(X_n)_{n \geq 1}$ is a subM w.r.t. filt $(\mathcal{F}_n)_{n \geq 1}$, then

\exists a closer for $(X_n)_{n \geq 1} \iff (X_n^+)_{n \geq 1}$ are UI

Proof:

(\Rightarrow) : Suppose (X_0, \mathcal{F}_0) closes $(X_n)_{n \geq 1}$.

$\therefore (X_0^+, \mathcal{F}_0)$ closes the subM $(X_n^+)_{n \geq 1}$ by trans of Ms.

$\therefore \underbrace{E(X_0^+ | \mathcal{F}_n)}_{\text{These are UI by proof of}} \stackrel{a.s.}{\geq} \underbrace{X_n^+}_{\text{Levy's thm}} \quad \forall n \in \mathbb{N}$

\therefore these are too.

\Leftrightarrow Suppose $(X_n^+)_n \geq 1$ are UI. (11)

$\therefore \exists X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$ s.t.

$$X_n \xrightarrow{\text{a.e.}} X_\infty$$

we show that $(X_\infty, \mathcal{F}_\infty)$ is a closer.

Case 1: $X_n \geq c > -\infty \quad \forall n \in \mathbb{N}$.

\therefore The X_n^- 's are UI & hence the X_n^+ 's are UI.

$\therefore X_n \xrightarrow{L_p} X_\infty$ by the L_p -convergence Thm.

Now to show $E(X_n | \mathcal{F}_n) \xrightarrow{\text{a.e.}} X_\infty$ Let

$A \in \mathcal{F}_n$ so that

$$\begin{aligned} \int_A X_n dP &\stackrel{\text{subM}}{\leq} \int_A E(X_{n+m} | \mathcal{F}_n) dP \\ &= \int_A X_{n+m} dP \\ &\xrightarrow{m \rightarrow \infty} \int_A X_\infty dP \quad \text{since } X_{n+m} \xrightarrow{L_p} X_\infty \text{ on } A \\ &= \int_A E(X_\infty | \mathcal{F}_n) dP \end{aligned}$$

$\therefore X_n \xrightarrow{\text{a.e.}} E(X_\infty | \mathcal{F}_n)$ by our results on indefinite integrals.

$\therefore (X_\infty, \mathcal{F}_\infty)$ closes $(X_n)_n \geq 1$.

Case 2: $X_n \in L_1(\Omega, \mathcal{F}, P)$.

Now case 1 applies to $\underbrace{X_n \vee c}_{\text{a subM since } \max(X_n, c)}$ where

$$c > -\infty.$$

a subM since $\max(X_n, c)$ is a subM, also

UI and bdd below.

$\therefore E(X_\infty \vee c | \mathcal{F}_n) \xrightarrow{\text{a.e.}} X_n \vee c$

$\therefore E(X_\infty | \mathcal{F}_n) \xrightarrow{\text{a.e.}} X_\infty$ by taking limits as $c \rightarrow -\infty$ by MCT for $E(\cdot | \mathcal{F}_n)$

QED

Now we get the "closer" condition for a.e. convergence almost as a corollary. (12)

Theorem: $(\exists \text{ closer} \Rightarrow \text{a.e. convergence})$

If the subM $(X_n)_n \geq 1$ has a closer

then $\exists X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$

s.t.

$$X_n \xrightarrow{\text{a.e.}} X_\infty$$

and $(X_\infty, \mathcal{F}_\infty)$ is the nearest closer.

Proof:

By the previous result

$\exists \text{ closer} \Leftrightarrow (X_n^+)_n$ are UI

$\Rightarrow E(X_n^+)_n$ are bdd

$\Rightarrow \exists X_\infty \in L_1(\Omega, \mathcal{F}_\infty, P)$ s.t

$$X_n \xrightarrow{\text{a.e.}} X_\infty.$$

The proof of the previous Thm also establishes that $(X_\infty, \mathcal{F}_\infty)$ closes $(X_n)_n \geq 1$.

Now we just show $(X_\infty, \mathcal{F}_\infty)$ is the nearest closer.

Let $(X_\bullet, \mathcal{F}_\bullet)$ be some closer.

We need to show

$X_1, X_2, \dots, X_\bullet, X_\infty$ is a subM

w.r.t. filtration $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_\bullet, \mathcal{F}_\infty$

it's a filtration
since $\cup_{i=1}^\infty \mathcal{F}_i = \mathcal{F}_\infty$

Notice it is sufficient to show since $\cup_{i=1}^\bullet \mathcal{F}_i = \mathcal{F}_\bullet$

$$(*) \quad E(X_\bullet | \mathcal{F}_\infty) \xrightarrow{\text{a.e.}} X_\infty$$

then $\cup_{i=1}^\bullet \mathcal{F}_i = \mathcal{F}_\bullet$

To show (*) use Lévy's smoothing result as (13) follows

$$E(X_n | \mathcal{F}_n) \stackrel{a.e.}{\geq} X_n \text{ since } (X_n, \mathcal{F}_n) \text{ is a closer}$$

$$\begin{array}{c} \downarrow a.e. \quad \downarrow a.e. \\ E(X_\infty | \mathcal{F}_\infty) \stackrel{a.e.}{\geq} X_\infty \end{array}$$

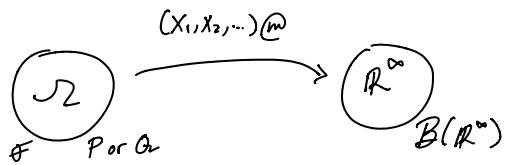
as way to be shown.

QED

Likelihood ratio Example from last Lecture

Two models P & Q for a random w generating an infinite sequence of r.v.s

$$X = (X_1, X_2, \dots)$$



where

P & Q are distinguishable (i.e.) $\Leftrightarrow P X^{-1} \perp Q X^{-1}$
from one sample of X

$$\Leftrightarrow Q_\infty \perp P_\infty$$

$$\text{where } \mathcal{F}_\infty := \sigma \langle X_i : i \geq 1 \rangle$$

$$Q_\infty := Q|_{\mathcal{F}_\infty}$$

$$\Leftrightarrow \frac{dQ_\infty}{dP_\infty} = 0$$

$$\Leftrightarrow \frac{dQ_\infty}{dP_n} \xrightarrow{a.e.} 0$$

$$\text{where } \mathcal{F}_n := \sigma \langle X_1, \dots, X_n \rangle$$

$$Q_n := Q|_{\mathcal{F}_n}$$

where $\frac{dQ_n}{dP_n}$ & $\frac{dQ_\infty}{dP_n}$ represents the finite & infinite data likelihood ratio.

Recall that we showed $\frac{dQ_n}{dP_n}$ is a sup M wrt $(\mathcal{F}_n)_{n \geq 1}$ under P

Theorem: $\frac{dQ_n}{dP_n} \xrightarrow{P-a.e.} \frac{dQ_\infty}{dP_\infty}$ as $n \rightarrow \infty$.

Proof:

$\frac{dQ_n}{dP_n}$ is a sup M and non-neg by construction

$\therefore \left(-\frac{dQ_n}{dP_n} \right)_{n \geq 1}$ is a non-positive subM

$\therefore (0, \mathcal{F}_\infty)$ closes $\left(-\frac{dQ_n}{dP_n} \right)_{n \geq 1}$ on the right:

$$\text{i.e. } -\frac{dQ_1}{dP_1}, -\frac{dQ_2}{dP_2}, \dots, 0$$

is a subM w.r.t. $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_\infty$ because

$$0 = E(0 | \mathcal{F}_n) \geq -\frac{dQ_n}{dP_n} \text{ a.s.}$$

By the Closer Thm $\exists X_\infty \in L(\mathcal{D}, \mathcal{F}_\infty, P)$ s.t.

$$\frac{dQ_n}{dP_n} \xrightarrow{a.e.} X_\infty \text{ & } (-X_\infty, \mathcal{F}_\infty) \text{ is the nearest closer of } \left(-\frac{dQ_n}{dP_n} \right)_{n \geq 1}.$$

$$\text{We show } X_\infty = \frac{dQ_\infty}{dP_\infty}$$

$$\left(\text{show } X_\infty \leq \frac{dQ_\infty}{dP_\infty} \right)$$

Use that $-X_\infty$ closes $\left(-\frac{dQ_n}{dP_n} \right)_{n \geq 1}$ on the right.
Since X_∞ closes we have

$$E(X_\infty | \mathcal{F}_n) \stackrel{P-a.e.}{\leq} \frac{dQ_n}{dP_n}$$

$$\therefore \int_A X_\infty dP \leq \int_A \frac{dQ_n}{dP_n} dP \leq Q(A) \quad \forall A \in \bigcup_{k=1}^\infty \mathcal{F}_k$$

Since $\bigcup_{k=1}^{\infty} \mathcal{F}_k$ is a field generating \mathcal{F}_{∞}
& $X_{\omega \geq 0}$ we can apply "Good sets" to
show

$$\underbrace{\int_A X_{\omega} dP}_{\text{a finite measure}} \leq Q_{\infty}(A) \quad \forall A \in \mathcal{F}_{\infty}$$

but $\frac{dQ_{\infty}^n}{dP_{\infty}}$ is the P -largest (by Leb Decp)

such, so we have

$$X_{\omega} \stackrel{\text{P-a.e.}}{\leq} \frac{dQ_{\infty}^n}{dP_{\infty}}$$

$$\left(\text{show } \underbrace{X_{\omega} \geq \frac{dQ_{\infty}^n}{dP_{\infty}}}_{\text{a.e.}} \right)$$

use that $-\frac{dQ_{\infty}^n}{dP_{\infty}}$ closes $(-\frac{dQ_n}{dP_n})_{n \geq 1}$ on the right.

Indeed this follows from the same method

used to show $(\frac{dQ_n}{dP_n})_{n \geq 1}$ is a supM.

nearest closer

$$\text{i.e. } -\frac{dQ_1}{dP_1}, -\frac{dQ_2}{dP_2}, \dots, \downarrow -X_{\omega}, -\frac{dQ_{\infty}^n}{dP_{\infty}}, 0$$

is a subM w.r.t. $\mathcal{F}_1, \mathcal{F}_2, \dots, \mathcal{F}_{\infty}, \mathcal{F}_{\infty}, \mathcal{F}_{\infty}$

$$\therefore E\left(-\frac{dQ_{\infty}^n}{dP_{\infty}} \mid \mathcal{F}_{\infty}\right) \stackrel{\text{a.e.}}{\geq} -X_{\omega}$$

// a.e.

$$-\frac{dQ_{\infty}^n}{dP_{\infty}}$$

\mathcal{QED}

Example : Polya's urn

L_p Convergence

Theorem: (subM L_p convergence Thm)

Suppose $1 \leq p < \infty$ and $(X_n)_{n \geq 1}$ is a subM

wrt filtration $(\mathcal{F}_n)_{n \geq 1}$.

If $|X_n|^p$ is UI then $\exists X_\infty \in L_p(\Omega, \mathcal{F}_\infty, P)$

s.t.

$$X_n \rightarrow X_\infty \quad \text{P-a.e. \& in } L_p$$

where $(X_\infty, \mathcal{F}_\infty)$ is the nearest closer of $(X_n)_{n \geq 1}$
 $\nwarrow = \sigma(\mathcal{F}_1, \mathcal{F}_2, \dots)$

Theorem: (Checking $|X_n|^p$ UI for $X_n \geq 0$ subM)

If $(X_n)_{n \geq 1}$ forms a non-neg subM & $p > 1$, then

$$X_n^p \text{ are UI} \Leftrightarrow \sup_n E(X_n^p) < \infty$$

$$\Leftrightarrow E\left(\sup_n X_n^p\right) < \infty$$