

## Lecture 23: 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.

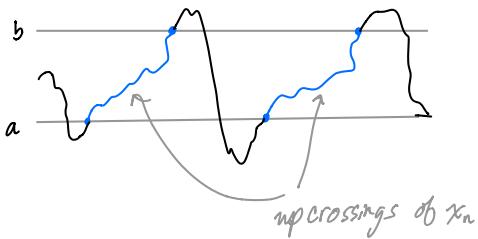
$x_n$  does not converge

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

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

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

i.e.



iff  $\exists a, b \in \mathbb{R}$  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

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

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

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

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

:

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

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

### 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  $\alpha_i$  is a ST.

To show  $\beta_i$  is a stopping time notice that

- If  $1 \leq m < n$  then

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

- 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  $\mathcal{F}_{\alpha_j}$ -measurable &

$X_{\beta_j}$  is  $\mathcal{F}_{\beta_j}$ -measurable

Since  $\mathcal{F}_{\alpha_j}$  &  $\mathcal{F}_{\beta_j}$  are sub  $\sigma$ -fields of  $\mathcal{F}_n$

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

QED.

Theorem: (Doob's upcrossing ineq)

(3)

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  $c > 0$  &  $\mathcal{U}_{0,c}$  denote the upcrossings of  $[0, c]$ . Then

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

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

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

must be  $X_1$  since  $X_1 > 0$

Notice that

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

Moreover

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

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

$$\begin{aligned} \therefore E(X_n - X_1) &\geq E(\underbrace{\text{blue}}_{\geq c \mathcal{U}_{0,c}}) + E(\underbrace{\text{red}}_{\geq 0}) \geq c E(\mathcal{U}_{0,c}). \\ &\quad \text{QED} \end{aligned}$$

Corollary:

If  $(X_1, \dots, X_n)$  is a subM w.r.t  $(\mathcal{F}_1, \dots, \mathcal{F}_n)$  and  $-\infty < a < b < \infty$  then

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

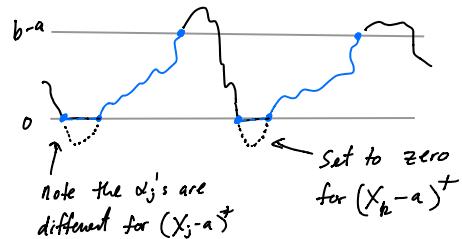
Proof:

For the first inequality:

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

upcrossings of  $[0, b-a]$  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

$$E(\mathcal{U}_{a,b}) \leq \frac{E(X_n - a)^+ - E(X_1 - a)^+}{b - a}$$

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:

$$\begin{aligned} \sup_n E(X_n^+) &\leq \underbrace{\sup_n E(X_n^+ I_{X_n^+ \geq c})}_{\xrightarrow{c \rightarrow \infty} 0 \text{ by defn of UI}} + \underbrace{\sup_n E(X_n^+ I_{X_n^+ < c})}_{< c} \\ &\quad \text{so for a large enough } c \text{ this is } < \infty. \end{aligned}$$

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  $X_\infty = E(X|\mathcal{F}_\infty)$ .

In particular show:

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

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

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

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

(8)

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  $(X_0, \mathcal{F}_0)$  <sup>r.v.</sup> <sup>sub σ-field of F</sup> 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:**  $(\exists \text{closer} \Leftrightarrow X_n^+ \text{'s are UI})$

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

$\exists$  a closer for  $(X_n)_{n \geq 1} \Leftrightarrow (X_n^+)_{n \geq 1}$  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.

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

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

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

$\Rightarrow E(X_n^+)'s \text{ are bdd}$

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

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

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

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

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

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

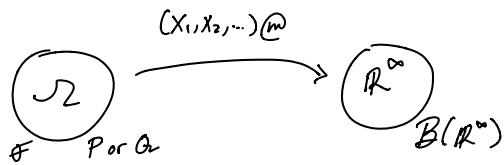
as way to be shown.

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  $X_\infty$  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  $\mathcal{F}_{\infty}$  &  $X_{n \geq 0}$  we can apply "Good sets" to show

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

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

such, so we have

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

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

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

Indeed this follows from the same method

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

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

nearest closer

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

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

// a.e.

$$\frac{dQ_{\infty}^a}{dP_{\infty}}$$

$\mathbb{Q} \in \mathcal{D}$

## L<sub>p</sub> Convergence

Theorem: (subM L<sub>p</sub> convergence Thm)

Suppose  $1 \leq p \leq \infty$  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  $(X_n)_{n \geq 1}$

$$\xrightarrow{\text{def}} = \sigma(\mathcal{F}_1, \mathcal{F}_2, \dots)$$

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

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

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