

1 a) A system \mathcal{V} (for all $\omega \in \Omega$)

is a collection of subsets of Ω so that

(i) $\Omega \in \mathcal{V}$

(ii) $A, B \in \mathcal{V} \Rightarrow A \cap B \in \mathcal{V}$

(iii) $A \in \mathcal{V} \Rightarrow A^c \in \mathcal{V}$

(b) Consider $X > 0$ a r.v. with $E(X) = \infty$

Take $X_n = \frac{X}{n}$ so $X_n \geq X_{n+1}$ and

$V_n E(X_n) = E(\frac{X}{n}) = \frac{1}{n} E(X) = \infty$. So that

$X_n = \frac{X}{n} \rightarrow 0$ so

$\lim_{n \rightarrow \infty} E(X_n) = \infty \neq E \lim_{n \rightarrow \infty} X_n = 0$.

(c) We wish to use the theorem $E = BC II$:

* If $V_n = \sum_{j=1}^n I_{A_j}$ where events A_j are pairwise disjoint

then $\sum_{j=1}^{\infty} P(A_j) \leq 1$

$\frac{E(V_n)}{V_n} = \frac{\sum_{j=1}^n P(A_j)}{\sum_{j=1}^n I_{A_j}}$

where $S_n = \sum_{j=1}^n I_{A_j}$ then $A_j = \{X_j > X_{j+1} + \frac{j}{2}\}$

(note difference with previous exam)

Now A_j, X_j, X_{j+1} are independent

$P(X_j, X_{j+1}) = P(X_j) P(X_{j+1})$ so $P(A_j) = P(X_j > X_{j+1} + \frac{j}{2})$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-y} e^{-x-y-\frac{j}{2}} dx dy = \int_{-\infty}^{\infty} e^{-2y-\frac{j}{2}} dy = \frac{1}{2} e^{-\frac{j}{2}}$$

$$(and so \sum_{j=1}^n P(A_j) = \sum_{j=1}^n \frac{1}{2j^{1/3}} \approx \frac{3}{2} n^{2/3} \rightarrow \infty)$$

hence the events $\{A_j\}_{j \in \mathbb{N}}$ are not pairwise

$$\text{indep. e.g. } P(A_k \cap A_{k+1}) \leq P(A_k \cap X_{k+1} > \frac{\log(k+1)}{3})$$

$$\leq P(X_k > \frac{\log k + \log(k+1)}{3}, X_{k+1} > \frac{\log(k+1)}{3})$$

$$\stackrel{\text{indep.}}{=} P(X_k > \frac{\log k + \log(k+1)}{3}) P(X_{k+1} > \frac{\log(k+1)}{3})$$

$$= \frac{1}{2^{(k+1)^{1/3}}} \frac{1}{(k+1)^{1/3}} \frac{1}{(k+1)^{1/3}} \ll P(A_k)P(A_{k+1})$$

For CLM

$$A \text{ trick is to use } S_n = \sum_{\substack{L \leq n \\ L \text{ odd}}} I_{A_L} + \sum_{\substack{L \leq n \\ L \text{ even}}} I_{A_L}$$

The events A_2, A_4, A_6, \dots are indep. same A_1, A_3, A_5, \dots

$$\text{so by } (*) \quad \frac{\sum_{\substack{L \leq n \\ L \text{ odd}}} I_{A_L}}{\sum_{\substack{L \leq n \\ L \text{ odd}}} P(A_L)} \rightarrow 1 \quad \text{and} \quad \frac{\sum_{\substack{L \leq n \\ L \text{ even}}} I_{A_L}}{\sum_{\substack{L \leq n \\ L \text{ even}}} P(A_L)} \rightarrow 1$$

$$\text{so } \frac{S_n}{E(S_n)} = \frac{\sum_{\substack{L \leq n \\ L \text{ odd}}} I_{A_L} + \sum_{\substack{L \leq n \\ L \text{ even}}} I_{A_L}}{\sum_{\substack{L \leq n \\ L \text{ odd}}} P(A_L) + \sum_{\substack{L \leq n \\ L \text{ even}}} P(A_L)} \rightarrow 1$$

$$\sum_{\substack{L \leq n \\ L \text{ odd}}} P(A_L) + \sum_{\substack{L \leq n \\ L \text{ even}}} P(A_L)$$

$$\text{so } \frac{S_n}{\frac{3}{2} n^{2/3}} \xrightarrow{\text{a.s.}} 1$$

$$d) \quad \overline{\lim} A_n = \bigcap_{n=1}^{\infty} \left(\bigcup_{m=n}^{\infty} A_m \right) \equiv \bigcap_{n=1}^{\infty} B_n$$

Now $B_n \downarrow \overline{\lim} A_n$ so by continuity of probs

$$P(\overline{\lim} A_n) = \lim_{n \rightarrow \infty} P(B_n) = \overline{\lim} P(B_n)$$

$$\geq \overline{\lim} P(A_n) \quad \text{since } \forall n \quad P(B_n) \geq P(A_n)$$

e) Fix $c > 0$ and consider $A_n = \{n^{1/2} X_n \leq c\}$
 $= \{X_n \leq c/n^{1/2}\}$. For $n^{1/2} \geq c$ $P(A_n) = \frac{c}{n^{1/2}}$.

So by BCT $P(\limsup A_n) = 1$ (as $\sum c/n^{1/2} = \infty$)

but on $\limsup A_n$ we have $\liminf n^{1/2} X_n \leq c$.

So as $\limsup X_n = 0$ as c is arbitrary > 0 and

$\forall n X_n > 0 \forall n$.

f) $\forall B P = Q$

show $\forall 0 \leq b-a, d-c \leq 1$

$[a, b] \cap [c, d] = [a \vee c, b \wedge d]$ is a clopen interval of length ≤ 1 , so

$\mathcal{T} = \{[a, b] : b-a \leq 1\}$ is a \mathcal{T} -system.

By Dynkin's \mathcal{T} - λ Lemma and our hypothesis on P, Q

$P(A) = Q(A) \forall A \in \sigma(\mathcal{T})$.

It remains to show

starting $\sigma(\mathcal{T}) = \mathcal{B}$:

For any $a < b \in \mathbb{R}$ $[a, b] = \bigcup_{n=1}^{\infty} [a, a + n \wedge b]$.

so $\{[a, b] : a < b\} \subseteq \sigma(\mathcal{T})$.

and $\exists a, b \in \mathbb{R} \quad \bigcup_{n=1}^{\infty} [a - \frac{1}{n}, b - \frac{1}{n}] \forall a < b$

so $\{]a, b[: a < b\} \subseteq \sigma(\mathcal{T})$. And any open set O

$= \bigcup I_i$ I_i is open interval, so

$O \subseteq \sigma(\mathcal{T}) \forall O$ open, so

$\mathcal{B} \subseteq \sigma(\mathcal{T})$ by defn of Bore σ -field

2a) X_n is X if $\forall \epsilon > 0 \quad P(|X_n - X| \geq \epsilon) \rightarrow 0$

(*) We use truncation. For a fixed $n \in \mathbb{N}$ (for $1 \leq c_n$)

$$X_i^c = X_i \wedge c_n \quad \text{for } c_n \text{ to be chosen.}$$

We write $S_n^c = \sum_{i=1}^n X_i^c$. We want c_n to be such

that $P(S_n^c \neq S_n) \rightarrow 0$. Subject to this we want

c_n "as small as possible".

$$P(S_n \neq S_n^c) \leq P(\exists c_n: X_i \neq X_i^c).$$

$$\sum_{i=1}^n P(X_i \neq X_i^c) = n P(X_1 \neq X_1^c) \quad (\text{i.i.d.})$$

$$= n \left(\frac{c_n}{c_n + 1} \right) \quad \text{if } c_n > 1 \quad (\text{we'll choose } c_n \text{ to be large})$$

We want $c_n = n \log n$ (log log n) then

$$n P(X_1 \neq X_1^c) = P(X_1 > c_n) = \frac{c_n}{n(c_n + 1)}$$

$$\leq \frac{\log \log n}{2} \leq \frac{n \log(n) + \log \log(n) + \log \log \log(n) + 1}{2}$$

for $n \rightarrow \infty$

So to show $\frac{2 S_n^c}{n \log n} \rightarrow 0$ it is enough to show that

$$\frac{2 S_n^c}{n \log n} \rightarrow 0$$

Now $E(S_n^c) = n E(X_1^c)$ or X_i^c is c_n if $c_n \leq X_i \leq c_n + 1$

$$= n \int_{c_n}^{\infty} P(X_1^c > x) dx = n + n \int_{c_n}^{\infty} P(X_1 > x) dx$$

$$= n + n \int_{c_n}^{\infty} \frac{1}{x} dx = n + n \log \frac{\infty}{c_n} = n + n \log n$$

$$= n + n \left(\log c_n + \log \frac{\infty}{c_n} \right) = n + n \log n$$

$$= n + n \log n + \frac{1}{2} n \log n + o(n)$$

26) CONTIN

SO TO PROVE $\frac{\sum S_n'}{n h^2 n} \xrightarrow{P} 1$ IT IS ENOUGH TO SHOW

$$\frac{S_n'}{E(S_n')} \xrightarrow{P} 1$$

so $\forall \epsilon > 0 \quad P\left(\left|\frac{S_n' - E(S_n')}{E(S_n')}\right| \geq \epsilon\right) \rightarrow 0.$

WE TAKE $\epsilon > 0$ WE WANT TO USE

$$(*) \quad P\left(\left|\frac{S_n' - E(S_n')}{E(S_n')}\right| \geq \epsilon\right) \leq \frac{1}{\epsilon^2 E(S_n')^2} \text{VAR}(S_n').$$

$$\stackrel{!}{=} \frac{n \text{VAR}(X_i')}{\epsilon^2 E(S_n')^2} \text{ AS } X_i' \text{ IS IID.}$$

$$\leq \frac{n E(X_i'^2)}{\epsilon^2 E(S_n')^2}$$

SO TO SHOW THIS REMAINS TO LEAD WE NEED ONLY GET A GOOD BOUND FOR $E(X_i'^2)$

$$E(X_i'^2) = \int_0^{\infty} 2x P(X_i' \geq x) dx$$

$$= \int_0^1 2x dx + \int_1^{C_n} 2x \frac{\log(C_n/x)}{2} dx$$

$$= 1 + 2 \int_1^{C_n} x \log(C_n/x) dx \leq 1 + 2C_n (\log(C_n) + 1)$$

$\leq 4 C_n \log n$ FOR ALL $n \leq 4 C_n \log n$ FOR ALL n

SO $\frac{n}{\epsilon^2 E(S_n')^2} E(X_i'^2) \leq \frac{16 n \log n \log n}{\epsilon^2 (n \log^2 n/h^2)}$

$$= \frac{24}{\epsilon^2} \frac{n^2 \log^2 n \log n}{n^2 \log^2 n} = \frac{24 \log n}{\epsilon^2 (\log n)^2} \xrightarrow{n \rightarrow \infty} 0$$

SO WE ARE DONE

(¹) NO! THE CONVERGENCE DOES NOT EXTEND TO
 C.S. WE ARGUE BY THE ABSURD!

SUPPOSE

$$\frac{2S_n}{n \ln^2 n} \xrightarrow{a.s.} 1. \text{ THEN } \frac{2S_{n-1}}{n \ln^2 n} = \frac{2S_{n-1}}{(n-1) \ln^2(n-1)} \cdot \frac{(n-1) \ln^2(n-1)}{n \ln^2(n)}$$

$$\rightarrow |x| = 1$$

$$\text{SO } \frac{2S_n}{n \ln^2 n} - \frac{2S_{n-1}}{n \ln^2 n} = \frac{2(S_n - S_{n-1})}{n \ln^2 n} \rightarrow | -1 | = 0$$

$$\text{AND } \frac{2S_n - S_{n-1}}{n \ln^2 n} = \frac{2X_n}{n \ln^2 n} \rightarrow \text{SUMS MUST TEND TO}$$

ZERO A.S.

BUT FOR EXAM $A_n = \{X_n > n \ln^2 n\}$ WE HAVE
 BY DEFINITION $X_n (= \text{sum of } X_i)$

$$P(A_n) = \frac{\ln(n \ln^2 n) + 1}{n \ln^2 n} \approx \frac{\ln n}{n \ln^2 n} = \frac{1}{n \ln n}$$

$$\frac{1}{n \ln n} \text{ FOR } n \gg 1$$

SO $\sum_{n=1}^{\infty} P(A_n) = \infty$ BY INTEGRAL TEST

SO AS X_n AND SO A_n ARE INDEPENDENT, WE MAY
 APPLY BC II TO CONCLUDE

$$\text{A.S. } \overline{\lim} \frac{2X_n}{n \ln^2 n} \geq 2. \text{ IN PARTICULAR}$$

$$\text{A.S. } \frac{2X_n}{n \ln^2 n} \not\rightarrow 0. \text{ CONTRADICTION, THEREFORE}$$

$$\frac{2S_n}{n \ln^2 n} \not\rightarrow 0$$

3) a) THIS IS STANDARD WAY TO PROVE EVERY ELEMENT OF A σ -FIELD HAS A GILMAN PROPERTY!

(i) SHOW EVERY ELEMENT OF A CONJUGATE GENERATING σ -COLLECTION HAS THE GILMAN PROPERTY

(ii) SHOW THAT THE SET OF SETS HAVING THE PROPERTY IS A σ -FIELD

FOCUS THE PROPERTY IS $W^{-1}(A) \in \mathcal{H}$ FOR $W: \Omega \rightarrow \mathcal{M}$ A GILMAN (OR P) FUNCTION

OUR σ -FIELD IS $\mathcal{G}(\mathcal{A})$ WHERE \mathcal{A} IS A COLLECTION OF SUBSETS OF \mathcal{M} SATISFYING THIS PROPERTY SO (i) ABOVE IS IMMEDIATE. FOR (ii) WE MUST SHOW THAT $\mathcal{G}(\mathcal{A}) \subseteq \mathcal{H}$: $W^{-1}(A) \in \mathcal{H}$ IS A σ -FIELD.

a) $\mathcal{H} \subseteq \mathcal{H}$: $W^{-1}(\mathcal{M}) = \Omega \in \mathcal{H}$ FOR ANY σ -FIELD

b) $A \in \mathcal{H} \Rightarrow A^c \in \mathcal{H}$: IF $A \in \mathcal{H} \Rightarrow W^{-1}(A) \in \mathcal{H} \Rightarrow W^{-1}(A)^c \in \mathcal{H}$. BUT $W^{-1}(A)^c = W^{-1}(A^c)$ SO $W^{-1}(A^c) \in \mathcal{H} \Rightarrow A^c \in \mathcal{H}$.

c) $A_n \in \mathcal{H} \forall n \Rightarrow \bigcup_n A_n \in \mathcal{H}$: $\forall_n A_n \in \mathcal{H} \Rightarrow \forall_n W^{-1}(A_n) \in \mathcal{H} \Rightarrow (\sigma$ -TRANSFORMED UNION CONJUGATE UNION) $\bigcup W^{-1}(A_n) \in \mathcal{H}$. BUT $\bigcup W^{-1}(A_n) = W^{-1}(\bigcup A_n)$ SO $W^{-1}(\bigcup A_n) \in \mathcal{H} \Rightarrow$ (PROP OF \mathcal{H}) $\bigcup A_n \in \mathcal{H}$.

(ii) TO SHOW THAT \mathcal{G} IS A PROB ON $(\mathcal{R}, \mathcal{F})$ WE MUST SHOW

a) $\mathcal{G}(\Omega) = 1$: $\mathcal{G}(\Omega) = E(X|_{\Omega}) = E(X) = 1$
(BY ADDITION OF X)

b) $\mathcal{G}(A) \in [0, 1] \forall A \in \mathcal{F}$: X IS ABSOLUTE SO $0 \leq E(X) \leq X$ ALSO BY MONOTONICITY OF THE INTEGRAL $0 \leq E(X) = \mathcal{G}(A) \leq E(X) = 1$.

$E(Y) = \sum_{i=1}^n p_i \cdot \mu_i$

Let $A = \begin{pmatrix} \mu_1 \\ \vdots \\ \mu_n \end{pmatrix}$
 $E(Y) = \sum_{i=1}^n p_i \mu_i$

$0 \leq \sum_{i=1}^n p_i \mu_i \leq \sum_{i=1}^n \mu_i$

So by Markov's inequality

$E(Y) = \sum_{i=1}^n p_i \mu_i = \sum_{i=1}^n p_i \mu_i$

(number of intervals plus $\sum_{i=1}^n p_i \mu_i$ as p_i is the number

$E(Y) = \sum_{i=1}^n p_i \mu_i = \sum_{i=1}^n p_i \mu_i$

$= \sum_{i=1}^n p_i \mu_i$

(ii) Let $\int_{-\infty}^{\infty} Y f(y) dy = E(Y)$. The error term

We use the "standard" approach to showing results for many bounded r.v.s.

a) For $Y = IA$

$E(Y) = E(IA) = \sum_{i=1}^n p_i \mu_i$

so results for Y of this form

for $Y = \sum_{i=1}^n c_i I_{A_i}$

By linearity $E(Y) = \sum_{i=1}^n c_i E(I_{A_i}) = \sum_{i=1}^n c_i p_i$

$E(Y) = \sum_{i=1}^n c_i p_i$

c) For Y general $\sum_{i=1}^n c_i I_{A_i}$

$Y = \sum_{i=1}^n c_i I_{A_i}$

then $E(Y) = \sum_{i=1}^n c_i E(I_{A_i}) = \sum_{i=1}^n c_i p_i$

where most bounds follow from P.M. cov. of (X, Y)

By Markov's concentration theorem

$$E(Y) = \sum_{i=1}^n E(Y_i) = E(X)$$

Since $Y_i \geq 0$, then

By Markov's concentration theorem: $A \geq \sum_{i=1}^n Y_i$

3) We may extend this to positive RVs Y