

MA 538 - Review session 1 for the Final

Problem 56. Let $X_1, X_2, \dots, X_n$, n be some real valued integrable random variables, independent and equally distributed. We set $m = \mathbf{E}[X_1]$ and $S_n = \sum_{i=1}^n X_i$.

56.1. Compute $\mathbf{E}[S_n|X_i]$ for all $i, 1 \leq i \leq n$.

$$\begin{aligned} E[S_n | X_i] &= E\left[\sum_{j=1}^n X_j | X_i\right] \\ &= \sum_{j=1}^n E[X_j | X_i] \\ &= E[X_i | X_i] + \sum_{j \neq i} E[X_j | X_i] \end{aligned}$$

Hence

$$E[S_n | X_i] = X_i + (n-1)m$$

56.2. Compute $E[X_i | S_n]$ for all $i, 1 \leq i \leq n$.

we use two facts:

$$(i) \quad \mathcal{L}(X_1, \dots, X_i, \dots, X_j, \dots, X_n) \\ = \mathcal{L}(X_1, \dots, X_j, \dots, X_i, \dots, X_n)$$

and

$$S_n(X_1, \dots, X_i, \dots, X_j, \dots, X_n) \\ = S_n(X_1, \dots, X_j, \dots, X_i, \dots, X_n)$$

Hence

$$E[X_i | S_n] = E[X_j | S_n]$$

$$(ii) \quad \sum_{i=1}^n E[X_i | S_n] = E[S_n | S_n] = S_n$$

Thus

$$E[X_i | S_n] = \frac{S_n}{n}$$

56.3. We assume now that $n = 2$ and that the random variables X_i have a common density f . Compute the conditional density of X_i given S_2 . Give a specific expression whenever the law of each X_i is an exponential law.

Density for the couple (X_1, S_2) Let $\varphi \in C_b(\mathbb{R}^2)$. We compute

$$\begin{aligned} \mathbb{E}[\varphi(X_1, S_2)] &= \mathbb{E}[\varphi(X_1, X_1 + X_2)] \\ &= \int_{\mathbb{R}^2} \varphi(x_1, x_1 + x_2) f(x_1) f(x_2) dx_1 dx_2 \end{aligned}$$

$$\text{CV: } x = x_1, \quad y = x_1 + x_2$$

$$\Rightarrow x_1 = x, \quad x_2 = y - x, \quad |J| = 1$$

Thus

$$\begin{aligned} \mathbb{E}[\varphi(X_1, S_2)] & \quad \text{density } f_{X_1, S_2} \text{ of couple } (X_1, S_2) \\ &= \int_{\mathbb{R}^2} \varphi(x, y) f(x) f(y-x) dx dy \quad (1) \end{aligned}$$

Case $X_i \sim \mathcal{E}(\lambda)$ Then the density of the couple (X_1, S_2) is, according to (1),

$$f_{X_1, S_2}(x, y) = \lambda e^{-\lambda x} \times \lambda e^{-\lambda(y-x)} \mathbb{1}_{(y > x > 0)}$$

$$\Rightarrow \boxed{f_{X_1, S_2}(x, y) = \lambda^2 e^{-\lambda y} \mathbb{1}_{(y > x > 0)}}$$

Conditional density According to our formula in class, we have

$$E[\varphi(x_1) | S_2] = h(S_2),$$

where

$$\begin{aligned} h(y) &= \frac{\int \varphi(x) f_{x, s_2}(x, y) dx}{\int f_{x, s_2}(x) dx} \\ &= \frac{\int \varphi(x) f(x) f(y-x) dx}{\int f(z) f(y-z) dz} \\ &= \int \varphi(x) f_{x|y}(x|y) dx, \end{aligned}$$

where the conditional density is

$$f_{x|y}(x|y) = \frac{f(x) f(y-x)}{\int f(z) f(y-z) dz}$$



Case $X_1 \sim E(\lambda)$ we have

$$f_{X_1|Y}(x|y) = \frac{f(x) f(y-x)}{\int f(z) f(y-z) dz}$$

Moreover

- $f(x) f(y-x) = \lambda^2 e^{-\lambda y} \mathbb{1}_{(0 < x < y)}$
- $\int f(z) f(y-z) dz$
 $= \int \lambda^2 e^{-\lambda y} \mathbb{1}_{(y > x > 0)} dx$
 $= \lambda^2 e^{-\lambda y} y$

Hence

$$f_{X_1|Y}(x|y) = \frac{1}{y} \mathbb{1}_{[0, y]}(x)$$



CRL we have obtained, in the $E(1)$ case, that

$$\mathcal{L}(X, 1S_2) = \mathcal{U}([0, S_2]) \equiv \mu(\omega, \cdot)$$

This is a CRL. That is

(i) For $\varphi \in C_b(\mathbb{R})$,

$$\begin{aligned}\mu(\omega, \varphi) &= \mathbb{E}[\varphi(X_1) | S_2] \\ &= \frac{1}{S_2(\omega)} \int_0^{S_2(\omega)} \varphi(x) dx\end{aligned}$$

Since $\omega \mapsto S_2(\omega)$ is measurable, we also have

$\omega \mapsto \mu(\omega, \varphi)$ measurable

(ii) For every $s \geq 0$, $\mathcal{U}([0, s])$ is a probability. Hence ω -a.s., $\mathcal{U}([0, S_2(\omega)])$ is a probability

Conclusion:

$\mathcal{U}([0, S_2])$ is a CRL

Problem 58. Let X and Y be two real valued random variables, such that Y follows an exponential law. We assume that given Y , X is distributed according to a Poisson law with parameter Y (given as a conditional regular law).

58.1. Compute the law of the couple (X, Y) , the law of X , and the law of Y given X as a conditional regular law.

Conditioning For $\varphi: \mathbb{N} \times \mathbb{R} \rightarrow \mathbb{R}$ we have

$$\begin{aligned} \mathbb{E}[\varphi(X, Y)] &= \mathbb{E}\left\{ \mathbb{E}[\varphi(X, Y) | Y] \right\} \\ &\stackrel{\text{crl}}{=} \mathbb{E}\left\{ \sum_{k=0}^{\infty} \varphi(k, Y) e^{-Y} \frac{Y^k}{k!} \right\}. \end{aligned}$$

Hence

$$\mathbb{E}[\varphi(X, Y)] = \sum_{k=0}^{\infty} \int_0^{\infty} \varphi(k, y) e^{-y} \frac{y^k}{k!} dy$$

Expression for the law we can write

$$\mathcal{L}(X, Y) = \sum_{k=0}^{\infty} \delta_k \otimes \mu_k,$$

where μ_k admits the density

$$f_k(y) = e^{-y} \frac{y^k}{k!} \mathbb{1}_{\mathbb{R}_+}(y) \quad (1)$$

computing $E[g(x)]$ From (1) we have

$$\begin{aligned} E[g(x)] &= \sum_{k=0}^{\infty} \int_0^{\infty} g(k) e^{-2y} \frac{y^k}{k!} dy \\ &= \sum_{k=0}^{\infty} \frac{g(k)}{k!} \int_0^{\infty} y^k e^{-2y} dy = I_k \end{aligned}$$

The quantity I_k can be computed thanks to a change of variable, i.e. letting $2y = z$ we get

$$\begin{aligned} I_k &= \int_0^{\infty} \left(\frac{z}{2}\right)^k e^{-z} \frac{dz}{2} \\ &= \frac{1}{2^{k+1}} \int_0^{\infty} z^k e^{-z} dz \\ &= \frac{1}{2^{k+1}} \Gamma(k+1) = \frac{k!}{2^{k+1}} \end{aligned}$$

Thus

$$E[g(x)] = \sum_{k=0}^{\infty} \frac{g(k)}{2^{k+1}}$$

Law of X Computing $\mathbb{E}[g(x)]$ for
 $g(k) = \mathbb{1}_{(k=n)}$, we get the pdf
of X :

$$P(X = n) = \frac{1}{2^{n+1}}, \quad n \geq 0$$

Note that $X \stackrel{(d)}{=} G - 1$, where
 $G \sim g(\frac{1}{2})$.

Computing $\mathbb{E}[h(Y)|X]$ since X is
a discrete random variable, we have

$$\mathbb{E}[h(Y)|X] = l(X),$$

where

$$l(n) = \frac{\mathbb{E}[h(Y) \mathbb{1}_{(X=n)}]}{P(X=n)}, \quad \text{for } n \geq 0$$

We thus compute, thanks to Q 1.1,

$$\begin{aligned} l(n) &= 2^{n+1} \sum_{k=0}^{\infty} \int_0^{\infty} h(y) \mathbb{1}_{(k=n)} e^{-2y} \frac{y^k}{k!} dy \\ &= \int_0^{\infty} h(y) \underbrace{2^{n+1} e^{-2y} \frac{y^n}{n!}}_{f_{Y|X}(y|n)} dy \end{aligned} \quad (1)$$

Problem 64. Let $\{Y_n; n \geq 1\}$ be a iid sequence of positive random variables such that $E[Y_j] = 1$, $P(Y_j = 1) < 1$ and $P(Y_j = 0) = 0$. We set

$$X_n = \prod_{j \leq n} Y_j.$$

64.1. Show that X is a martingale.

Filtration we take

$$\mathcal{F}_n = \sigma(Y_1, \dots, Y_n)$$

Integrability we have seen

$$z, w \in L^1(\Omega), \quad z \perp w \Rightarrow zw \in L^1(\Omega)$$

Here since $Y_j \in L^1(\Omega)$ and the Y_j 's are independent, we get $X_n \in L^1(\Omega)$

Martingale property we have

$$\begin{aligned} E[X_{n+1} | \mathcal{F}_n] &= E[X_n Y_{n+1} | \mathcal{F}_n] \\ &= X_n E[Y_{n+1} | \mathcal{F}_n] \quad (X_n \in \mathcal{F}_n) \\ &= X_n E[Y_{n+1}] \quad (Y_{n+1} \perp \mathcal{F}_n) \\ &= X_n \end{aligned}$$

Hence (X_n) is a martingale

64.2. Show that $\lim_{n \rightarrow \infty} X_n = 0$ a.s

Convergence we have

X_n martingale, $X_n \geq 0$

$\Rightarrow X_n \rightarrow X_\infty$ a.s., with $X_\infty \in L^1(\Omega)$

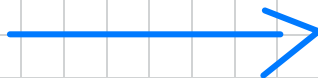
we still need to show that $X_\infty = 0$.

Log-sequence set

$$Z_n = \ln(X_n)$$

Then

$$Z_n = \sum_{j=1}^n \ln(Y_j)$$



Easy case we have

$$\mathbb{E}[\ln(Y_i)] \leq \ln(\mathbb{E}[Y_i]) \stackrel{\text{Hyp}}{=} 0$$

Assume

$$\ln(Y_i) \in L^1, \quad -m \equiv \mathbb{E}[\ln(Y_i)] < 0$$

Then by LLN we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} z_n = -m \quad \text{a.s.}$$

$$\Rightarrow \lim_{n \rightarrow \infty} z_n = -\infty \quad \text{a.s.}$$

$$\Rightarrow \lim_{n \rightarrow \infty} x_n = 0 \quad \text{a.s.}$$

General strategy we have seen

that $x_n \rightarrow x_\infty$ a.s., with $x_\infty \in [0, \infty)$ a.s. For $z_n = \ln(x_n)$ this means

$$\lim_{n \rightarrow \infty} z_n = \begin{cases} -\infty & \text{if } x_\infty = 0 \\ z_\infty \in (-\infty, \infty) & \text{if } x_\infty > 0 \end{cases}$$

If we wish to prove $x_\infty = 0$, we just have to show that

$$z_\infty = \sum \ln(Y_i) \quad \text{divergent}$$

Proof that $\sum \ln(Y_i)$ divergent It is sufficient to prove that

$\ln(Y_i)$ does not converge to 0

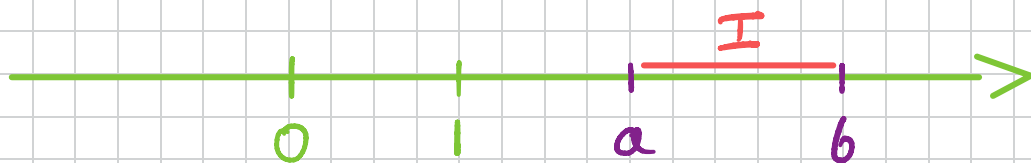
This is due to the fact that
" Y_i is away from 1 i.o."

Bounding Y_i Since

$$P(Y_i = 1) < 1, \quad E[Y_i] = 1$$

there exists $I = [a, b]$ with $a > 1$ s.t.

$$P(Y_i \in I) = p > 0$$



$$\text{Thus } \sum_{j=1}^n P(Y_j \in I) = \infty$$

Inverse B-C

$$\Rightarrow P(Y_i \in I \text{ i.o.}) = 1$$

$$\Rightarrow \boxed{Z_\infty = 0}$$

Problem 69. Let $(Y_n)_{n \geq 1}$ be a sequence of independent random variables with a common law given by $\mathbf{P}(Y_n = 1) = p = 1 - \mathbf{P}(Y_n = -1) = 1 - q$. We define $(S_n)_{n \in \mathbb{N}}$ by $S_0 = 0$ and $S_n = \sum_{k=1}^n Y_k$.

69.1. We assume that $p = q = \frac{1}{2}$. We set $T_a = \inf\{n \geq 0, S_n = a\}$ ($a \in \mathbb{Z}^*$). Show that $\mathbf{E}(T_a) = +\infty$.

setting of Pb 68 S_n is a martingale satisfying the conditions of Problem 68, i.e.

$$|X_n - X_{n-1}| \leq 1 \equiv M$$

If $\nu \equiv T_a$ also fulfills the conditions of Pb 68, that is

$$T_a \in L'(\Omega),$$

then we would get

$$\mathbf{E}[S_{T_a}] = \mathbf{E}[0] = 0 \quad (1)$$

contradiction Relation (1) is impossible:

$$S_{T_a} = a \text{ a.s.}, \text{ hence } \mathbf{E}[S_{T_a}] = a$$

Thus

$$T_a \notin L'(\Omega)$$

69.2. Let $T = T_{a,b} = \inf\{n \geq 0, S_n = -a \text{ or } S_n = b\}$ ($a, b \in \mathbb{N}$). Using the value of $E(S_T)$, compute the probability of the event $(S_T = -a)$.

T is a.s. finite If $x \in (-a, b)$
we have

$$\mathbb{P}(x + S_{a+b} \notin (-a, b)) \geq \frac{1}{2^{b+a}} > 0$$

Then write

$$\begin{aligned} a_{n+1} &\equiv \mathbb{P}(T > (n+1)(a+b)) \\ &= E\left\{ \mathbb{1}_{(T > n(a+b))} \mathbb{1}_{(S_{n(a+b)} + \sum_{k=1}^{a+b} Y_{n(a+b)+k} \in (-a, b))} \right\} \\ &= E\left\{ \mathbb{1}_{(T > n(a+b))} E\left[\mathbb{1}_{(S_{n(a+b)} + \tilde{S}_{a+b} \in (-a, b))} \mid \mathcal{F}_{n(a+b)} \right] \right\} \\ &= E\left[\mathbb{1}_{(x + \tilde{S}_{a+b} \in (-a, b))} \mid x = S_{n(a+b)} \right] \\ &\leq \sup_{x \in (-a, b)} \mathbb{P}(x + \tilde{S}_{a+b} \in (-a, b)) \\ &\leq 1 - \frac{1}{2^{b+a}} \equiv \rho \end{aligned}$$

Hence

$$a_{n+1} \leq \rho E[\mathbb{1}_{(T > n(a+b))}] = \rho a_n$$

we get $a_n \leq a_0 \rho^n$, hence
 $\lim_{n \rightarrow \infty} a_n = 0$ and

$T < \infty$ a.s.

Decomposing S_T we have

$$S_T = -a \mathbb{1}_{(S_T = -a)} + b \mathbb{1}_{(S_T = b)}$$

Optional stopping we have that

$$\{S_{n \wedge T}; n \geq 1\}$$

is a bounded martingale. Hence

$$\mathbb{E}[S_T] = 0$$

which can be written as

$$-a \underbrace{\mathbb{P}(S_T = -a)}_x + b \underbrace{\mathbb{P}(S_T = b)}_y = 0$$

System The couple (x, y) solves

$$\begin{cases} x + y = 1 \\ -ax + by = 0 \end{cases}$$

$$\Rightarrow (a+b)y = a$$

$$\Rightarrow y = \frac{a}{a+b} \quad x = \frac{b}{a+b}$$

we get

$$\mathbb{P}(S_T = -a) = \frac{b}{a+b}$$

69.3. Show that $Z_n = S_n^2 - n$ is a martingale, and from the value of $\mathbf{E}(Z_T)$ compute $\mathbf{E}(T)$.

z_n is a martingale we have

(i) $z_n = \varphi_n(S_n) \Rightarrow z_n \in \mathcal{F}_n$

(ii) $|S_n| \leq n$, hence $z_n \in L^\infty(\Omega)$
 $\Rightarrow z_n \in L^1(\Omega)$

(iii) we have

$$\begin{aligned} z_{n+1} &= S_{n+1}^2 - (n+1) \\ &= (S_n + Y_{n+1})^2 - (n+1) \\ &= S_n^2 + 2S_n Y_{n+1} + \overbrace{Y_{n+1}^2}^{=1} - (n+1) \\ &= z_n + 2S_n Y_{n+1} \end{aligned}$$

Hence

$$\begin{aligned} \mathbb{E}[z_{n+1} | \mathcal{F}_n] &= z_n + 2S_n \mathbb{E}[Y_{n+1} | \mathcal{F}_n] \\ &= z_n + 2S_n \mathbb{E}[Y_{n+1}] = z_n \end{aligned}$$

hence

(z_n) is a martingale

Optional stopping problem Here T is not bounded and

$\{Z_{n \wedge T} ; n \geq 0\}$ not bounded

Application of the martingale prop For a fixed n we have

$$\mathbb{E}[Z_{n \wedge T}] = 0$$

$$\Rightarrow \mathbb{E}[S_{n \wedge T}^2] = \mathbb{E}[T \wedge n] \quad (2)$$

Limit in lhs of (1) we have

$$S_{n \wedge T} \xrightarrow{a.s.} S_T, \text{ since } T < \infty \text{ a.s.}$$

$$|S_{n \wedge T}^2| \leq \max(a^2, b^2)$$

By dominated convergence,

$$\lim_{n \rightarrow \infty} \mathbb{E}[S_{n \wedge T}^2] = \mathbb{E}[S_T^2]$$

Limit in lhs of (1) By monotone convergence

$$\lim_{n \rightarrow \infty} \mathbb{E}[T \wedge n] = \mathbb{E}[T]$$

Computing $E[T]$ we have obtained

$$E[T] = E[S_T^2]$$

$$= a^2 P(S_T = -a) + b^2 P(S_T = b)$$

$$= \frac{a^2 b}{a+b} + \frac{a b^2}{a+b}$$

Hence

$$E[T] = ab$$