

Problem Sheet 1

1. Let $\Omega = \{1, 2, 3\}$. Let

$$\begin{aligned}\mathcal{F} &= \{\emptyset, \{1\}, \{2, 3\}, \{1, 2, 3\}\}, \\ \mathcal{F}' &= \{\emptyset, \{2\}, \{1, 3\}, \{1, 2, 3\}\}.\end{aligned}$$

You may assume that both \mathcal{F} and \mathcal{F}' are σ -fields.

- (a) Show that $\mathcal{F} \cup \mathcal{F}'$ is not a σ -field.
 (b) Let $X : \Omega \rightarrow \mathbb{R}$ be defined by

$$X(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2 & \text{if } n = 2 \\ 1 & \text{if } n = 3 \end{cases}$$

Is X measurable with respect to \mathcal{F} ? Is X measurable with respect to \mathcal{F}' ?

2. Let Ω be any set. Let I be any set and for each $i \in I$ let \mathcal{F}_i be a σ -field on Ω . Prove that $\bigcap_{i \in I} \mathcal{F}_i$ is a σ -field on Ω .
3. Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space and let $X : \Omega \rightarrow \mathbb{R}$ be a random variable.
- (a) Show that X^2 is a random variable.
 (b) Let $g : \mathbb{R} \rightarrow \mathbb{R}$ be such that $g(x) \geq g(y)$ for all $x \geq y$. Show that $g(X)$ is a random variable.

4. Let $\Omega = \{0, 1\}^{\mathbb{N}}$, and let us write each $\omega \in \Omega$ as a sequence: $\omega = \omega_1 \omega_2 \omega_3 \dots$ where $\omega_i \in \{0, 1\}$. Let $\mathcal{F} = \sigma(\{\omega; \omega_n = C\}; n \in \mathbb{N}, C \in \{0, 1\})$. For each $n \in \mathbb{N}$ let $X_n : \Omega \rightarrow \mathbb{R}$ be given by $X_n(\omega) = \omega_n$ and define

$$S_n = \sum_{i=1}^n X_i.$$

- (a) Show that the following events are \mathcal{F} measurable:

$$\{\forall n, X_n = 1\}, \quad \{\exists N, \forall n \geq N, X_n = 0\}, \quad \left\{ \sup_{m \leq n} S_m \leq \frac{n}{2} \right\}.$$

Suppose additionally that $\mathbb{P} : \mathcal{F} \rightarrow [0, 1]$ is a probability measure, under which the X_i are independent and identically distributed with $\mathbb{P}[X_1 = 0] = \mathbb{P}[X_1 = 1] = \frac{1}{2}$.

- (c) Calculate $\mathbb{E}[S_2 | \sigma(X_1)]$ and $\mathbb{E}[S_2^2 | \sigma(X_1)]$.
 (d) Let $n \in \mathbb{N}$. Calculate $\mathbb{E}[X_1 | \sigma(S_n)]$.
5. Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space and let $X \in L^1$. Let \mathcal{G} be a sub- σ -field of \mathcal{F} .
- (a) Prove that $\mathbb{E}[\mathbb{E}[X | \mathcal{G}]] = \mathbb{E}[X]$ almost surely.
 (b) Let $\mathcal{F}_0 = \{\emptyset, \Omega\}$. Prove that there exists $c \in \mathbb{R}$ such that $\mathbb{E}[X | \mathcal{F}_0] = c$ almost surely. Hence, show that $c = \mathbb{E}[X]$.
6. Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space and let $X, Y \in L^1$. Let \mathcal{G} be a sub- σ -field of \mathcal{F} . Suppose that $\mathbb{E}[X | \mathcal{G}] = Y$ and $\mathbb{E}[X^2] = \mathbb{E}[Y^2]$. Prove that $X = Y$ almost surely.

Problem Sheet 2

1. Let $(X_n)_{n \in \mathbb{N}}$ be an iid sequence of random variables such that $\mathbb{P}[X_1 = -1] = \mathbb{P}[X_1 = 1] = \frac{1}{2}$. Let

$$S_n = \sum_{i=1}^n X_i.$$

Let $\mathcal{F}_n = \sigma(X_i; i \leq n)$.

- (a) Show that \mathcal{F}_n is a filtration and that S_n is a \mathcal{F}_n martingale.
 (b) State, with proof, which of the following processes are \mathcal{F}_n martingales:

$$(i) S_n^2 \quad (ii) S_n^2 - n \quad (iii) \frac{S_n}{n}$$

Which of the above are submartingales?

2. Let X_0, X_1, \dots be a sequence of \mathcal{L}^1 random variables. Let \mathcal{F}_n be a filtration and suppose that $\mathbb{E}[X_{n+1} | \mathcal{F}_n] = aX_n + bX_{n-1}$ for all $n \in \mathbb{N}$, where $a, b > 0$ and $a + b = 1$.

Find a value of $\alpha \in \mathbb{R}$ for which $S_n = \alpha X_n + X_{n-1}$ is an \mathcal{F}_n martingale.

3. At time 0, an urn contains 1 black ball and 1 white ball. At each time $n = 1, 2, 3, \dots$, a ball is chosen from the urn and returned to the urn. At the same time, a new ball of the same colour as the chosen ball is added to the urn. Just after time n , there are $n + 2$ balls in the urn, of which $B_n + 1$ are black, where B_n is the number of black balls added into the urn at or before time n .

Let

$$M_n = \frac{B_n + 1}{n + 2}$$

be the proportion of balls in the urn that are black, at time n . Note that $M_n \in [0, 1]$.

- (a) Show that, relative to a natural filtration that you should specify, M_n is a martingale.
 (b) Calculate the probability that the first k balls drawn are all black and that the next j balls drawn are all white.
 (c) Show that $\mathbb{P}[B_n = k] = \frac{1}{n+1}$ for all $0 \leq k \leq n$, and deduce that $\lim_{n \rightarrow \infty} \mathbb{P}[M_n \leq p] = p$ for all $p \in [0, 1]$.
 (d) Let T be the number of balls drawn until the first black ball appears. Show that T is a stopping time and use the Optional Stopping Theorem to show that $\mathbb{E}[\frac{1}{T+2}] = \frac{1}{4}$.
4. Let S and T be stopping times with respect to the filtration \mathcal{F}_n .
- (a) Show that $\min(S, T)$ and $\max(S, T)$ are stopping times.
 (b) Suppose $S \leq T$. Is it necessarily true that $T - S$ is a stopping time?
5. Suppose that we repeatedly toss a fair coin, writing H for heads and T for tails. What is the expected number of tosses until we have seen the pattern $HTHT$ for the first time?
 Give an example of a four letter pattern of $\{H, T\}$ that has the maximal expected number of tosses, of any four letter pattern, until it is seen.

6. Let $m \in \mathbb{N}$ and $m \geq 2$. At time $n = 0$, an urn contains $2m$ balls, of which m are red and m are blue. At each time $n = 1, \dots, 2m$ we draw a single ball from the urn; we do not replace it. Therefore, at time n the urn contains $2m - n$ balls.

Let N_n denote the number of red balls remaining in the urn at time n . For $n = 0, \dots, 2m - 1$ let

$$P_n = \frac{N_n}{2m - n}$$

be the fraction of red balls remaining after time n . Let $\mathcal{G}_n = \sigma(N_i; i \leq n)$.

- (a) Show that P_n is a \mathcal{G}_n martingale.
 (b) Let T be the first time at which the ball that we draw is red. Note that $T < 2m$, because the urn initially contains at least 2 red balls. Show that the probability that the $(T + 1)^{st}$ ball is red is $\frac{1}{2}$.

Problem Sheet 3

1. You play a game by betting on outcome of i.i.d. random variables X_n , $n \in \mathbb{Z}^+$, where

$$\mathbb{P}[X_n = 1] = p, \quad \mathbb{P}[X_n = -1] = q = 1 - p, \quad \frac{1}{2} < p < 1.$$

Let Z_n be your fortune at time n , that is $Z_n = Z_0 + \sum_{j=1}^n C_j X_j$. The bet C_n you place on game n must be in $(0, Z_{n-1})$ (i.e. you cannot borrow money to place bets). Your objective is to maximise the expected ‘interest rate’ $\mathbb{E}[\log(Z_N/Z_0)]$, where N (the length of the game) and Z_0 (your initial fortune) are both fixed. Let $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$. Show that if C is a previsible strategy, then $\log Z_n - n\alpha$ is a supermartingale, where

$$\alpha = p \log p + q \log q + \log 2,$$

and deduce that $\mathbb{E} \log[Z_n/Z_0] \leq N\alpha$.

Can you find a strategy such that $\log Z_n - n\alpha$ is a martingale?

2. Let \mathcal{F}_n be a filtration. Suppose T is a stopping time such that for some $K \geq 1$ and $\epsilon > 0$ we have, for all $n \geq 0$, almost surely

$$\mathbb{P}[T \leq n + K \mid \mathcal{F}_n] \geq \epsilon.$$

- (a) Prove by induction that for all $m \in \mathbb{N}$, $\mathbb{P}[T \geq mK] \leq (1 - \epsilon)^m$.
 (b) Hence show that $\mathbb{E}[T] < \infty$.

3. Let X_1, X_2, \dots be a sequence of iid random variables with

$$\mathbb{P}[X_1 = 1] = p, \quad \mathbb{P}[X_1 = -1] = q, \quad \text{where } 0 < p = 1 - q < 1,$$

and suppose that $p \neq q$. Let $a, b \in \mathbb{N}$ with $0 < a < b$, and let

$$S_n = a + X_1 + \dots + X_n, \quad T = \inf\{n \geq 0; S_n = 0 \text{ or } S_n = b\}.$$

Let $\mathcal{F}_n = \sigma(X_i; i \leq n)$.

- (a) Deduce from the previous question that $\mathbb{E}[T] < \infty$.
 (b) Show that

$$M_n = \left(\frac{q}{p}\right)^{S_n}, \quad N_n = S_n - n(p - q)$$

are both \mathcal{F}_n martingales.

- (c) Calculate $\mathbb{P}[S_T = 0]$, $\mathbb{E}[S_T]$ and hence calculate $\mathbb{E}[T]$.

4. Let X_1, X_2, \dots be strictly positive iid random variables such that $\mathbb{E}[X_1] = 1$ and $\mathbb{P}[X_1 = 1] < 1$.

- (a) Show that $M_n = \prod_{i=1}^n X_i$ is a martingale relative to a natural filtration that you should specify.
 (b) Deduce that there exists a real valued random variable L such that $M_n \rightarrow L$ almost surely as $n \rightarrow \infty$.
 (c) Show that $\mathbb{P}[L = 0] = 1$.
Hint: Argue by contradiction and note that if $M_n, M_{n+1} \in (c - \epsilon, c + \epsilon)$ then $X_{n+1} \in (\frac{c - \epsilon}{c + \epsilon}, \frac{c + \epsilon}{c - \epsilon})$.
 (d) Use the Strong Law of Large Numbers to show that there exists $c \in \mathbb{R}$ such that $\frac{1}{n} \log M_n \rightarrow c$ almost surely $n \rightarrow \infty$. Use Jensen’s inequality to show that $c < 0$.

5. Show that a set \mathcal{C} of random variables is uniformly integrable if either:

- (a) There exists a random variable Y such that $\mathbb{E}[|Y|] < \infty$ and $|X| \leq Y$ for all $X \in \mathcal{C}$.
 (b) There exists $p > 1$ and $A < \infty$ such that $\mathbb{E}[|X|^p] \leq A$ for all $X \in \mathcal{C}$.

6. Let Z_n be a Galton-Watson process with offspring distribution G (which takes value in $0, 1, \dots$), where $\mathbb{E}[G] = \mu > 1$ and $\text{var}[G] = \sigma^2 < \infty$. Set $M_n = \frac{Z_n}{\mu^n}$, and use the filtration from lecture notes.

Show that M_n is a martingale. Find a formula $\mathbb{E}[M_n^2]$ in terms of n, μ and σ . Hence, show that $\sup_{n \in \mathbb{N}} \mathbb{E}[M_n^2] < \infty$ and that M_n converges almost surely and in L^1 as $n \rightarrow \infty$. Deduce that the limit M_∞ satisfies $\mathbb{P}[M_\infty > 0] > 0$.

Problem Sheet 2

Q3

$B_n = \# \text{black at time } n$

$$M_n = \frac{B_n + 1}{n + 2}$$

(a) Take

$$\tilde{\mathcal{F}}_n = \sigma(B_i : i \leq n).$$

Then $\tilde{\mathcal{F}}_n \subseteq \tilde{\mathcal{F}}_{n+1}$ so $\tilde{\mathcal{F}}$ is a filtration. Since sums & products of measurable functions are measurable, and deterministic functions are measurable, and $B_n \in m\sigma(B_n) \subseteq m\tilde{\mathcal{F}}_n$,

$$M_n \in m\tilde{\mathcal{F}}_n.$$

We have $M_n \in [0, 1)$ so $M_n \in L'$

Given \mathcal{F}_n , the probability of picking a black ball at time $n+1$ is $\frac{B_{n+1}}{n+2}$

If a black ball is picked then

$$M_{n+1} = \frac{B_{n+2}}{n+3}, \quad \text{if not then } M_n = \frac{B_{n+1}}{n+3}.$$

Hence,

$$\begin{aligned} & E[M_{n+1} | \mathcal{F}_n] \\ &= \left(\frac{B_{n+1}}{n+2} \right) \left(\frac{B_{n+2}}{n+3} \right) + \left(1 - \frac{B_{n+1}}{n+2} \right) \left(\frac{B_{n+1}}{n+3} \right) \\ &= \left(\frac{B_n^2 + 3B_n + 2}{(n+2)(n+3)} \right) + \left(\frac{(n+1 - B_n)(B_{n+1})}{(n+2)(n+3)} \right) \\ &= \left(\frac{B_n^2 + 3B_n + 2}{(n+2)(n+3)} \right) + \left(\frac{-B_n^2 + nB_n + (n+1)}{(n+2)(n+3)} \right) \\ &= \frac{B_n(n+3) + (n+3)}{(n+2)(n+3)} \\ &= \frac{B_n + 1}{n+2} = M_n. \end{aligned}$$

Hence M_n is a martingale.

(b) The probability of drawing k black balls followed by j white is

$$\underbrace{\left(\frac{1}{2}\right)\left(\frac{2}{3}\right)\dots\left(\frac{k}{k+1}\right)}_{k \text{ black}} \times \underbrace{\left(\frac{1}{k+2}\right)\left(\frac{2}{k+3}\right)\dots\left(\frac{j}{k+j+1}\right)}_{j \text{ white}} = \frac{k! j!}{(k+j+1)!}$$

(c) Note that, in (b), any other way in which we can draw k black balls and j white balls has the same probability - because the denominator stays the same & the numerator is permuted. Hence,

$$P[B_n = k] = \binom{n}{k} \frac{k! (n-k)!}{(n+1)!} = \frac{1}{n+1}$$

Note that

$$\mathbb{P}(N_n \leq p)$$

$$= \mathbb{P}(B_n \leq p(n+2) - 1)$$

$$= \sum_{k=1}^{\lfloor p(n+2) - 1 \rfloor} \mathbb{P}(B_n = k)$$

$$= \lfloor p(n+2) - 1 \rfloor \frac{1}{n+1}$$

$\rightarrow p$ as $n \rightarrow \infty$.

(d) Note that

$$\{T=n\} = \left(\bigcap_{i=1}^{n-1} \{B_i=0\} \right) \cap \{B_n=1\}$$

since $B_i \in \mathcal{F}_n$ for all $i \leq n$, we have $\{T=n\} \in \mathcal{F}_n$.

Hence T is a stopping time.

M_n is bounded &

$$\begin{aligned} \mathbb{P}(\tau = \infty) &= \mathbb{P}(\forall n, B_n = 0) \\ &= \left(\frac{1}{2}\right)\left(\frac{1}{3}\right)\left(\frac{1}{4}\right)\dots \\ &= 0 \end{aligned}$$

Hence, the Optional Stopping Theorem applies and

$$\mathbb{E}[M_\tau] = \mathbb{E}[M_0] = \frac{1}{2}.$$

We have

$$\begin{aligned} \mathbb{E}[M_\tau] &= \mathbb{E}\left[\frac{B_\tau + 1}{\tau + 2}\right] \\ &= \mathbb{E}\left[\frac{2}{\tau + 2}\right] \end{aligned}$$

Hence,

$$\mathbb{E}\left[\frac{1}{\tau + 2}\right] = \frac{1}{4}.$$

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$N_n = \# \text{ red balls at time } n$

$$P_n = \frac{N_n}{2^{m-n}}$$

(a) Since $N_n \in \mathcal{G}_n$, deterministic
functions are measurable & products
of measurable functions are measurable,
we have $P_n \in \mathcal{G}_n$.

Since $0 \leq N_n \leq 2^{m-n}$ we have
 $|P_n| \leq 1$, hence $P_n \in L^1$.

Given \mathcal{G}_n , the probability that
the n^{th} draw is red is P_n .

If the n^{th} draw is red then

$$P_{n+1} = \frac{N_n - 1}{2^{m-n-1}}, \text{ if not then } P_n = \frac{N_n}{2^{m-n-1}}$$

Hence,

$$E(P_{atn} | \mathcal{F}_n)$$

$$= P_n \left(\frac{N_n - 1}{2m - n - 1} \right) + (1 - P_n) \left(\frac{N_n}{2m - n - 1} \right)$$

$$= \frac{N_n^2 - N_n + (2m - n)N_n - N_n^2}{(2m - n)(2m - n - 1)}$$

$$= \frac{N_n(2m - n - 1)}{(2m - n)(2m - n - 1)} = P_n.$$

Hence P_n is a martingale.

(b) Since

$$\{T = n\} = \left(\bigcap_{i=0}^{n-1} \{N_i = n\} \right) \cap \{N_n = n-1\}$$

& $N_i \in \mathcal{m}\mathcal{F}_n$ for all $i \leq n$, we have

$\{T = n\} \in \mathcal{m}\mathcal{F}_n$, hence T is a stopping time.

Moreover, $T \leq 2m$ so T is a bounded stopping time. Hence the Optional Stopping Theorem applies and

$$\mathbb{E}[P_T] = \mathbb{E}[P_0] = 1/2.$$

Hence,

$$\begin{aligned} \mathbb{P}((T+1)^{\text{st}} \text{ is red}) &= \mathbb{P}(N_{T+1} = N_T + 1) \\ &= \sum_{i=1}^{2m-1} \mathbb{P}(N_{T+1} = N_T + 1 \mid T=i) \mathbb{P}(T=i) \\ &= \sum_{i=1}^{2m-1} \frac{m-i}{2m-i} \mathbb{P}(T=i) \\ &= \mathbb{E}[P_T] = 1/2. \end{aligned}$$

2) Let $K \geq 1$, $\varepsilon > 0$.

$$P[T \leq n + K | \mathcal{F}_n] \geq \varepsilon.$$

(a) For $m = 0$,

$$P[T \geq 0] \leq (1 - \varepsilon)^0 = 1$$

Now consider $m \geq 1$,

$$P[T \geq mK]$$

$$= E[\mathbb{1}(T \geq mK) \mathbb{1}(T \geq (m-1)K)]$$

$$= E\left[E[\mathbb{1}(T \geq (m-1)K) \mathbb{1}(T \geq mK) | \mathcal{F}_{(m-1)K}]\right]$$

$$= E\left[\mathbb{1}(T \geq (m-1)K) \underbrace{E[\mathbb{1}(T \geq mK) | \mathcal{F}_{(m-1)K}]}_{\leq (1-\varepsilon)}\right]$$

$$\text{set } n = (m-1)K$$

$$\& \leq (1 - \varepsilon)$$

$$\leq (1 - \varepsilon) E[\mathbb{1}(T \geq (m-1)K)]$$

It follows by induction that

$$P(T \geq mk) \leq (1-\varepsilon)^m.$$

(b) Noting that

$$E[T] = \sum_{n=0}^{\infty} P(T \geq n),$$

we have

$$\begin{aligned} E[T] &\leq \sum_{n=0}^{\infty} \sum_{n=mk}^{(n+1)k-1} P(T \geq mk) \\ &\leq \sum_{n=0}^{\infty} k(1-\varepsilon)^{\lfloor n/k \rfloor} < \infty. \end{aligned}$$

$$3) S_n = a + X_1 + \dots + X_n$$

$$P[X_i = 1] = p \quad P[X_i = -1] = q$$

$$p \neq q$$

$$0 < a < b.$$

$$T = \inf \{n \geq 0 : S_n = 0 \text{ or } S_n = b\}.$$

(a) For all n ,

$$P[X_i = 1 \text{ for all } i = n+1, \dots, n+b] \\ = p^b > 0$$

If we step upwards b time in a row, starting at $S_n > 0$, then $S_{n+b} \geq b$.

Hence,

$$P[T \geq n+b \mid \mathcal{F}_n] \geq p^b$$

$$\text{By Q2, } E[T] < \infty.$$

(b) I leave it for you to show that M_n, N_n are $\mathbb{E} \mathcal{F}_n$ and \mathcal{L}' .

$$\mathbb{E}[N_{n+1} | \mathcal{F}_n]$$

$$= \mathbb{E}[S_{n+1} | \mathcal{F}_n] - (n+1)(p-q)$$

$$= S_n + \mathbb{E}[X_{n+1} | \mathcal{F}_n] - (n+1)(p-q)$$

$$= S_n + (p-q) - (n+1)(p-q) = N_n.$$

$$\mathbb{E}[M_{n+1} | \mathcal{F}_n]$$

$$= \left(\frac{q}{p}\right)^{S_n} \mathbb{E}\left[\left(\frac{q}{p}\right)^{X_{n+1}} | \mathcal{F}_n\right]$$

$$= M_n \mathbb{E}\left[\left(\frac{q}{p}\right)^{X_{n+1}}\right]$$

$$= M_n \left(\left(\frac{q}{p}\right)^1 p + \left(\frac{q}{p}\right)^{-1} q \right)$$

$$= M_n$$

Here, we use the taking out what is known rule & the relationship between conditional expectation & independence.

(c) Since $\mathbb{E}[T] < \infty$, we may apply the Optional Stopping Theorem at T .

Hence,

$$\mathbb{E}[S_T - T(p-q)] = \mathbb{E}[S_0] = a$$

$$\mathbb{E}[S_T] = a + \mathbb{E}[T](p-q).$$

$$\mathbb{E}\left[\left(\frac{q}{p}\right)^{S_T}\right] = \mathbb{E}\left[\left(\frac{q}{p}\right)^{S_0}\right] = \left(\frac{q}{p}\right)^a$$

Hence,

$$\left(\frac{q}{p}\right)^a = 1 \cdot \mathbb{P}[S_T = 0] + \left(\frac{q}{p}\right)^b \mathbb{P}[S_T = b]$$

Since $\mathbb{E}[T] < \infty$, $\mathbb{P}[T < \infty] = 1$.

Hence $\mathbb{P}[S_T = 0] + \mathbb{P}[S_T = b] = 1$. So,

$$\left(\frac{q}{p}\right)^a - \left(\frac{q}{p}\right)^b = \mathbb{P}[S_T = 0] \left(1 - \left(\frac{q}{p}\right)^b\right)$$

So

$$\mathbb{P}[S_T = 0] = \frac{\left(\frac{q}{p}\right)^a - \left(\frac{q}{p}\right)^b}{1 - \left(\frac{q}{p}\right)^b}.$$

Hence,

$$\mathbb{P}[S_T = b] = 1 - \frac{(q/p)^a - (q/p)^b}{1 - (q/p)^b}$$

$$= \frac{1 - (q/p)^a}{1 - (q/p)^b}$$

So

$$\mathbb{E}[S_T] = b \left(\frac{1 - (q/p)^a}{1 - (q/p)^b} \right)$$

$$\mathbb{E}[\tau] = \frac{b \left(\frac{1 - (q/p)^a}{1 - (q/p)^b} \right) - a}{p - q}$$

Q5

(a) Let $\varepsilon > 0$.

Since $Y \in L^1$, $\mathbb{E}[|Y| \mathbb{1}_{(|Y| \geq K)}] \rightarrow 0$
as $K \rightarrow \infty$.

Hence, choose K s.t.

$$\mathbb{E}[|Y| \mathbb{1}_{(|Y| \geq K)}] < \varepsilon.$$

Then, for any $x \in C$,

$$\begin{aligned} \mathbb{E}[|x| \mathbb{1}_{(|x| \geq K)}] \\ \leq \mathbb{E}[|Y| \mathbb{1}_{(|Y| \geq K)}] < \varepsilon. \end{aligned}$$

So C is UI.

(b) Note that

$$\begin{aligned} \mathbb{E}[|x| \mathbb{1}_{(|x| \geq K)}] \\ \leq \frac{1}{K^{p-1}} \mathbb{E}[K^{p-1} |x| \mathbb{1}_{(|x| \geq K)}] \\ \leq K^{1-p} \mathbb{E}[|x|^p] \leq AK^{1-p} \end{aligned}$$

Let $\varepsilon > 0$. Choose K s.t. $AK^{1-p} < \varepsilon$.

Then $\mathbb{E}[|x| \mathbb{1}_{(|x| \geq K)}] < \varepsilon$.

Hence C is UI.

Q6

$$\mathbb{E}(G) = \mu > 1 \quad \text{var}(G) = \sigma^2 < \infty$$

Z_n is Galton process, offspring distⁿ G .

$$Z_1 = 1.$$

We have

$$Z_{n+1} = \sum_{i=1}^{Z_n} X_{n+1,i} \quad \oplus$$

where $(X_{n,i})$ are iid with distⁿ G .

Let

$$\mathcal{F}_n = \sigma(X_{m,i} : i \in \mathbb{N}, m \leq n).$$

Then $Z_1 = 1$ so $Z_1 \in m \mathcal{F}_1$.

If $Z_n \in m \mathcal{F}_n$ then $Z_n \in m \mathcal{F}_{n+1}$, so

by \oplus we have $Z_{n+1} \in m \mathcal{F}_{n+1}$.

(Since sums of meas. func^s are meas.).

Hence, by induction, $Z_n \in m \mathcal{F}_n$
for all n .

We note,

$$\mathbb{E}[Z_{n+1} | \mathcal{F}_n]$$

$$= \sum_{k=1}^{\infty} \mathbb{E}[Z_{n+1} \mathbb{1}(Z_n = k) | \mathcal{F}_n]$$

$$= \sum_{k=1}^{\infty} \mathbb{E}[X_{n+1,1} + \dots + X_{n+1,k} | \mathcal{F}_n] \mathbb{1}(Z_n = k)$$

$$= \sum_{k=1}^{\infty} \mathbb{1}(Z_n = k) \mathbb{E}[X_{n+1,1} + \dots + X_{n+1,k}]$$

$$= \sum_{k=1}^{\infty} \mathbb{1}(Z_n = k) k \mathbb{E}[G]$$

$$= \mu Z_n$$

$$\text{Hence, } \mathbb{E}[Z_n / \mu^{n+1} | \mathcal{F}_n] = \frac{Z_n}{\mu^n}.$$

So M_n is a martingale.

We note that

$$\begin{aligned} \mathbb{E}[(M_{n+1} - M_n)^2 | \mathcal{F}_n] \\ &= \mathbb{E}[M_{n+1}^2 | \mathcal{F}_n] - 2M_n \mathbb{E}[M_{n+1} | \mathcal{F}_n] \\ &\quad + M_n^2 \\ &= \mathbb{E}[M_{n+1}^2 | \mathcal{F}_n] - M_n^2. \end{aligned}$$

Here, we use the taking out what is known rule & the martingale property of M . Hence,

$$\mathbb{E}[M_{n+1}^2 | \mathcal{F}_n] = M_n^2 + \mathbb{E}[(M_{n+1} - M_n)^2 | \mathcal{F}_n].$$

We want to use induction to find a formula for $\mathbb{E}[M_{n+1}^2]$, so we try to calculate the final term.

$$\mathbb{E}[(M_{n+1} - M_n)^2 | \mathcal{F}_n]$$

$$= \mu^{-2(n+1)} \mathbb{E}[(z_{n+1} - \mu z_n)^2 | \mathcal{F}_n]$$

and

$$\mathbb{E}[(z_{n+1} - \mu z_n)^2 | \mathcal{F}_n]$$

$$= \sum_{k=1}^{\infty} \mathbb{E}[(z_{n+1} - \mu z_n)^2 \mathbb{1}(z_n = k) | \mathcal{F}_n]$$

$$= \sum_{k=1}^{\infty} \mathbb{1}(z_n = k) \mathbb{E}[(X_{n+1,1} + \dots + X_{n+1,k} - \mu k)^2 | \mathcal{F}_n]$$

$$= \sum_{k=1}^{\infty} \mathbb{1}(z_n = k) \mathbb{E}[(X_{n+1,1} + \dots + X_{n+1,k} - \mu k)^2]$$

$$= \sum_{k=1}^{\infty} \mathbb{1}(z_n = k) \mathbb{E}[(Y_1 + \dots + Y_k)^2]$$

where $Y_i = X_{n+1,i} - \mu$.

Note that the (Y_i) are independent and that $\mathbb{E}[Y_i] = 0$.

$$\text{Also, } \mathbb{E}[Y_i^2] = \text{var}[X_{n+1,i}] = \sigma^2$$

Hence,

$$\begin{aligned}\mathbb{E}[(Y_1 + \dots + Y_k)^2] \\ &= \sum_{i=1}^k \mathbb{E}[Y_i^2] \\ &= k\sigma^2\end{aligned}$$

Hence,

$$\begin{aligned}\mathbb{E}[(Z_{n+1} - \mu Z_n)^2 \mid \mathcal{F}_n] \\ &= \sum_{k=1}^{\infty} \mathbb{1}(Z_n = k) k\sigma^2 \\ &= Z_n \sigma^2.\end{aligned}$$

So

$$\mathbb{E}[(M_{n+1} - M_n)^2 \mid \mathcal{F}_n] = \frac{Z_n \sigma^2}{n^{2(n+1)}}.$$

Hence,

$$\mathbb{E}[M_{n+1}^2] = \mathbb{E}[M_n^2] + \frac{\mathbb{E}[Z_n] \sigma^2}{\mu^{2(n+1)}}$$

$$= \mathbb{E}[M_n^2] + \frac{\mu^{n-1} \sigma^2}{\mu^{2(n+1)}}$$

$$= \mathbb{E}[M_n^2] + \frac{\sigma^2}{\mu^{n+3}}$$

Therefore,

$$\mathbb{E}[M_{n+1}^2] \leq \sum_{i=1}^n \frac{\sigma^2}{\mu^{i+3}} \leq \sum_{i=1}^{\infty} \frac{\sigma^2}{\mu^{i+3}} < \infty.$$

$$\text{So } \sup_n \mathbb{E}[M_n^2] < \infty.$$

Hence also, (M_n) is uniformly integrable.

By the martingale convergence theorem,

$\exists M_\infty$ s.t. $M_n \rightarrow M_\infty$ both almost

surely and in L^1 . Thus $\mathbb{E}[M_n] \rightarrow \mathbb{E}[M_\infty]$.

Hence $\mathbb{E}[M_\infty] = \frac{1}{\mu} > 0$.

Since $M_n \geq 0$, $M_\infty \geq 0$.

Hence, $\mathbb{P}[M_\infty > 0] > 0$.