

Problem Sheet 4

1. X_n be the position after n steps of a particle that performs an unbiased random walk on \mathbb{Z} starting at 0.

Show that $(X_n^2 - n)_{n \geq 0}$ is a martingale w.r.t $(X_n)_{n \geq 0}$

define $(Y_n)_{n \geq 0} = (X_n^2 - n)_{n \geq 0}$. $(Y_n)_{n \geq 0}$ is a martingale if the following three conditions hold.

i) $E(|Y_n|) < \infty$

This is great!

$$E(|X_n^2 - n|) \leq E(|X_n^2|) + E(|n|) : \text{Triangle inequality}$$

$$= E(X_n^2) + E(n) : X_n^2 \text{ and } n \geq 0 \therefore \text{no need for the modulus}$$

$$= E(X_n^2) + n : \text{expectation of a constant is a constant}$$

$\max(X_n^2) = n^2$
 $\therefore \dots \leq n^2 + n$

$$= E\left(\left(\sum_{i=1}^n \zeta_i\right)^2\right) + n : X_n = \sum_{i=1}^n \zeta_i$$

where $\zeta_i = \pm 1$ w.p. $\frac{1}{2}$

$$= E\left(\underbrace{\sum_{i=1}^n \zeta_i^2}_{(1)} + \underbrace{\sum_{i \neq j} \zeta_i \zeta_j}_{(2)}\right) + n$$

(1) represents all times where both ζ 's are equal to 1 or -1

(2) represents all times where one $\zeta = 1$ and the other $\zeta = -1$

$$= \sum_{i=1}^n E(\zeta_i^2) + \sum_{i \neq j} E(\zeta_i \zeta_j) + n$$

$$= \sum_{i=1}^n \cdot 1 + \sum_{i \neq j} \cdot 0 + n$$

where $E(\zeta_i^2) = (1)^2 \left(\frac{1}{2}\right) + (-1)^2 \left(\frac{1}{2}\right) = 1$

and $E(\zeta_i \zeta_j) = \underbrace{E(\zeta_i)E(\zeta_j)}_{\text{independence}} = \underbrace{\left((1)\left(\frac{1}{2}\right) + (-1)\left(\frac{1}{2}\right)\right)}_{\text{+10}} \left((-1)\left(\frac{1}{2}\right) + (1)\left(\frac{1}{2}\right)\right) = 0$

$$= n + n \leq 2n < \infty \quad \square$$

$$\text{ii) } E(Y_n | X_{n-1}, \dots, X_0) = Y_{n-1}$$

$$E(X_n^2 - n | X_{n-1}, \dots, X_0) = E(X_n^2 | X_{n-1}, \dots, X_0) - E(n | X_{n-1}, \dots, X_0);$$

$$X_n^2 = (X_{n-1} + \zeta_n)^2 \Rightarrow \overset{\text{linearity}}{=} E((X_{n-1} + \zeta_n)^2 | X_{n-1}, \dots, X_0) - n : \\ \text{independence and expectation of a constant}$$

now expand the quadratic

$$= \underbrace{E(X_{n-1}^2 | X_{n-1}, \dots, X_0)}_{\text{measurable}} + 2 \underbrace{E(X_{n-1} \zeta_n | X_{n-1}, \dots, X_0)}_{\text{measurable}} + \underbrace{E(\zeta_n^2 | X_{n-1}, \dots, X_0)}_{\text{independent}} - n$$

$$= X_{n-1}^2 + 2X_{n-1}E(\zeta_n) + E(\zeta_n^2) - n$$

from i) we know that $E(\zeta_n) = 0$ and $E(\zeta_n^2) = 1$

$$\begin{aligned} \Rightarrow &= X_{n-1}^2 + 2X_{n-1}(0) + 1 - n \\ &= X_{n-1}^2 - n + 1 \\ &= X_{n-1}^2 - (n-1) \\ &= Y_{n-1} \end{aligned}$$

□ ✓+14

iii) $Y_n = X_n^2 - n$ is adapted to $(X_n)_{n \in \mathbb{N}_0}$ because Y_n is a function of X_n . □

The three conditions hold, and therefore $(Y_n)_{n \in \mathbb{N}_0} = (X_n^2 - n)_{n \in \mathbb{N}_0}$ is a martingale w.r.t. $(X_n)_{n \geq 0}$ ✓+5

2. X_n represents an unbiased random walk and is the same as the random variable X_n seen in the first question.

Show that $P(|X_n| \geq t) \leq 2e^{-\frac{t^2}{2n}}$

We will use Hoeffding's inequality.

The Hoeffding inequality states that a martingale $(X_n)_{n \in \mathbb{N}}$ satisfies the bounded difference condition, i.e. for any $n \geq 1$

$$P(|X_n - X_{n-1}| \leq k_n) = 1 \quad \text{for some } k_n > 0$$

$$P(|X_n - X_0| \geq x) \leq 2e^{-\frac{1}{2} \cdot \frac{x^2}{\sum_{i=1}^n k_i^2}}$$

If X_n represents an unbiased random walk: $X_n = \begin{cases} X_{n-1} + 1 & \text{w.p. } \frac{1}{2} \\ X_{n-1} - 1 & \text{w.p. } \frac{1}{2} \end{cases}$
where $X_0 = 0$ starting at 0 then:

i) we know that this X_n is a martingale (proven in lecture notes)

ii) $|X_n - X_{n-1}| = 1 = k_n \Rightarrow P(|X_n - X_{n-1}| = 1) = 1 \quad \forall n \geq 1$ ✓+5
 \Rightarrow the bounded difference condition is satisfied

iii) $X_0 = 0$ ✓+5

Using Hoeffding's inequality, setting $x = t$ and substituting the above properties:

$$P(|X_n - 0| \geq t) \leq 2 \cdot e^{-\frac{1}{2} \cdot \frac{t^2}{\sum_{i=1}^n 1^2}}$$

$$\Rightarrow P(|X_n| \geq t) \leq 2 \cdot e^{-\frac{t^2}{2n}} \quad \text{✓+5}$$

$$\text{Since } \sum_{i=1}^n 1^2 = \sum_{i=1}^n 1 = n$$

□

3. Z_n : size of n^{th} generation ; $Z_0=1$; $E(Z) < 1$

show that $\lim_{n \rightarrow \infty} P(Z_n > 0) = 0$

$$P(Z_n > 0) = P(Z_n \geq 1) \leq E(Z_n) \quad \checkmark : \text{markov inequality; } P(X \geq E) \leq \frac{E(X)}{E}$$

~~not~~ $(Z_n)_{n \geq 0}$ is not a martingale, however, we can rescale.

let $W_n = \frac{Z_n}{\mu^n}$ / where $(W_n)_{n \geq 0}$ is a martingale w.r.t $(Z_n)_{n \geq 0}$ (proven in lecture notes)

Since $(W_n)_{n \geq 0}$ is a martingale, $E(W_n) = E(W_0)$ / $\forall n \geq 0$

$$\Rightarrow E\left(\frac{Z_n}{\mu^n}\right) = E\left(\frac{Z_0}{\mu^0}\right) = E(Z_0) = 1 \quad \checkmark$$

$$E\left(\frac{Z_n}{\mu^n}\right) = 1 \Rightarrow E(Z_n) \frac{1}{\mu^n} = 1 \quad \checkmark$$

$E(Z_n) = \mu^n$ ~~o~~

Substitute ① into the inequality above.

$$P(Z_n > 0) \leq \mu^n \quad \checkmark$$

Take the limit of both sides

$$\lim_{n \rightarrow \infty} P(Z_n > 0) \leq \lim_{n \rightarrow \infty} \mu^n \quad \checkmark$$

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if $E(Z_n) < 1 \forall n$ then $\lim_{n \rightarrow \infty} \mu^n = 0$

$$\lim_{n \rightarrow \infty} P(Z_n > 0) \leq 0 \Rightarrow \lim_{n \rightarrow \infty} P(Z_n > 0) = 0 \quad \checkmark$$

a probability cannot be less than zero

~~very nice.~~ \square

4.) X_n : biased random walk $X_n = \begin{cases} X_{n-1} + 1 & \text{w.p. } p \\ X_{n-1} - 1 & \text{w.p. } q = 1-p \end{cases}$
 where $0 < p < 1$.

Show that ~~the~~ $(Y_n)_{n \geq 0}$, where $Y_n = \left(\frac{q}{p}\right)^{X_n}$ is a martingale with respect to $(X_n)_{n \geq 0}$.

~~Let~~ define $X_n = \sum_{i=1}^n Z_i$ where $Z_i = \begin{cases} +1 & \text{w.p. } p \\ -1 & \text{w.p. } q = 1-p \end{cases}$

the following three conditions must hold for $(Y_n)_{n \geq 0}$ to be a martingale w.r.t. $(X_n)_{n \geq 0}$

i) $E(|Y_n|) < \infty$

$$E(|Y_n|) = E\left|\left(\frac{q}{p}\right)^{X_n}\right|$$

$$= E\left(\left(\frac{q}{p}\right)^{X_n}\right) \quad \text{if } 0 < p < 1 \text{ then } \frac{q}{p} > 0 \text{ and } \left(\frac{q}{p}\right)^{X_n} > 0 \quad \forall X_n \in \mathbb{Z}$$

$$= E\left(\left(\frac{q}{p}\right)^{Z_1 + Z_2 + \dots + Z_n}\right)$$

$$= E\left(\left(\frac{q}{p}\right)^{Z_1} \left(\frac{q}{p}\right)^{Z_2} \dots \left(\frac{q}{p}\right)^{Z_n}\right)$$

$$= E\left(\left(\frac{q}{p}\right)^{Z_1}\right) \cdot E\left(\left(\frac{q}{p}\right)^{Z_2}\right) \dots E\left(\left(\frac{q}{p}\right)^{Z_n}\right) \quad \checkmark$$

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

$$= \frac{q}{p} \cdot p + \frac{p}{q} \cdot q$$

$$= q + p$$

$$= 1$$

$$\Rightarrow E(|Y_n|) = \prod_{i=1}^n E\left(\left(\frac{q}{p}\right)^{Z_i}\right) = \prod_{i=1}^n 1 = 1 < \infty \quad \square$$

$$\begin{aligned}
\text{ii) } E(Y_n | X_{n-1}, \dots, X_0) &= Y_{n-1} \\
E((a/p)^{X_n} | X_{n-1}, \dots, X_0) &= E((a/p)^{\sum_{i=1}^n Z_i} | X_{n-1}, \dots, X_0) \\
&= E((a/p)^{Z_n + \sum_{i=1}^{n-1} Z_i} | X_{n-1}, \dots, X_0) \\
&= E((a/p)^{Z_n} (a/p)^{\sum_{i=1}^{n-1} Z_i} | X_{n-1}, \dots, X_0) \\
&= \underbrace{E((a/p)^{Z_n} | X_{n-1}, \dots, X_0)}_{\text{independence}} \underbrace{E((a/p)^{\sum_{i=1}^{n-1} Z_i} | X_{n-1}, \dots, X_0)}_{\text{measurable}} \\
&= E((a/p)^{Z_n}) \cdot (a/p)^{\sum_{i=1}^{n-1} Z_i}
\end{aligned}$$

From part i) we have shown that $E((a/p)^{Z_i}) = 1$

$$\begin{aligned}
\Rightarrow E(Y_n | X_{n-1}, \dots, X_0) &= 1 \cdot (a/p)^{\sum_{i=1}^{n-1} Z_i} \\
&= (a/p)^{X_{n-1}} \\
&= Y_{n-1} \quad \square
\end{aligned}$$

iii) $Y_n = (a/p)^{X_n}$ is adapted to $(X_n)_{n \in N_0}$, because Y_n is a function of X_n .

The three conditions hold and therefore $(Y_n)_{n \in N_0}$ is a martingale w.r.t. $(X_n)_{n \in N_0}$. \square

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