

Assignment - 1

- Let $M(t)$ be a symmetric random walk. Check if $M(t)^2 - t$ is a martingale with respect to filtration $\mathcal{F}_t = \sigma(X_1, X_2, \dots, X_t)$.

→ As $M(t)$ is a symmetric random walk.

$$M(t) = \sum_{j=1}^t X_j, \quad X_j \begin{cases} +1 & p=0.5 \\ -1 & p=0.5 \end{cases}$$

a) $M(t) = \sum_{j=1}^t X_j$ is \mathcal{F}_t -measurable since it depends only on $X_j, j \leq t$, i.e. on the information available at time t .

t is not a random variable & it also depends on information available at time t . Hence t is also \mathcal{F}_t -measurable.

$\therefore M(t)^2 - t$ only depends on information available at time t , Hence $M(t)^2 - t$ is \mathcal{F}_t measurable.

b) As $M(t) = \sum_{j=1}^t X_j$

$$|M(t)| = \left| \sum_{j=1}^t X_j \right|$$

$$|M(t)| \leq \sum_{j=1}^t |X_j| \quad \left[\text{As } |a+b| \leq |a|+|b| \right]$$

The fact: $\therefore |M(t)| \leq t, \quad t \geq 1, |M(t)| \geq 0$

Squaring both sides

$$|M(t)|^2 \leq t^2$$

but as $|a^2| = |a|^2$

$$\therefore |M(t)^2| \leq t^2 \quad \text{--- (1)}$$

Now,

$$|M(t)^2 - t| \leq |M(t)^2| + |-t|$$

as $t \geq 0$

$$\therefore |M(t)^2 - t| \leq |M(t)^2| + t$$

Using (1)

$$|M(t)^2 - t| \leq t^2 + t$$

$$\therefore E[|M(t)^2 - t|] \leq t^2 + t < \infty$$

$$\therefore E[|M(t)^2 - t|] < \infty$$

$$\begin{aligned}
 c) E(M(t)^2 - t | F_s) &= E((M(t) - M(s))^2 + 2M(t)M(s) - M(s)^2 - t | F_s) \\
 &= E((M(t) - M(s))^2 | F_s) + 2E[M(t)M(s) | F_s] - E[M(s)^2 | F_s] - t
 \end{aligned}$$

At time s , $M(s)$ is observed & no longer random. Also $M(t) - M(s)$ does not depend on info available at time s .

$$\begin{aligned}
 \therefore E(M(t)^2 - t | F_s) &= E[(M(t) - M(s))^2] + 2M(s)E[M(t) | F_s] - M(s)^2 - t \\
 &= \quad \quad \quad - (2)
 \end{aligned}$$

As $M(t)$ is a symmetric random walk & we have proved that a symmetric random walk is a martingale (in class notes)

$$\therefore E[M(t) | F_s] = M(s)$$

$$\begin{aligned}
 \text{As } E[(M(t) - M(s))^2] &= \text{Var}[M(t) - M(s)] \\
 &= \text{Var}\left[\sum_{j=s+1}^t X_j\right] \\
 &= \sum_{j=s+1}^t \text{Var}(X_j) = t - s \quad \text{as } \text{Var}(X_j) = 1
 \end{aligned}$$

$\therefore E_1^{(2)}$ becomes

$$E(M(t)^2 - t | F_s) = t - s + 2M(s) \cdot M(s) - M(s)^2 - t$$

$$\therefore E(M(t)^2 - t | \mathcal{F}_s) = M(s)^2 + S$$

Which proves part (c) of martingality property

As $M(t)^2 - t$ satisfies all 3 properties of martingality $M(t)^2 - t$ is a martingale.

2. Compute $E e^{\sigma M(t)}$ for a given constant $\sigma > 0$ and fixed t .

$$\rightarrow E[e^{\sigma M(t)}] = E\left[e^{\sigma \sum_{j=1}^t X_j}\right]$$

$$= E\left[\prod_{j=1}^t e^{\sigma X_j}\right]$$

As X_j 's are i.i.d. so are $e^{\sigma X_j}$'s

$$\therefore E[e^{\sigma M(t)}] = \prod_{j=1}^t E[e^{\sigma X_j}]$$

$$\begin{aligned} E[e^{\sigma X_j}] &= \frac{1}{2} e^{\sigma} + \frac{1}{2} e^{-\sigma} \\ &= \frac{1}{2} \left[e^{\sigma} + \frac{1}{e^{\sigma}} \right] = \frac{1}{2} \left[\frac{e^{2\sigma} + 1}{e^{\sigma}} \right] \end{aligned}$$

$$\therefore E[e^{\sigma M(t)}] = \prod_{j=1}^t \frac{1}{2} \left[\frac{e^{2\sigma} + 1}{e^{\sigma}} \right]$$

$$\therefore E[e^{\sigma M(t)}] = \left(\frac{1}{2} \left[\frac{e^{2\sigma} + 1}{e^{\sigma}} \right] \right)^t$$

3. Give an example of random Variable X such that $E|X| = \infty$

→ Consider a discrete random Variable X that takes on the value 2^n with probability $\frac{1}{2^n}$ for $n=1,2,3,\dots$

$$\text{i.e. } P(X=2^n) = \frac{1}{2^n} \text{ for } n=1,2,3,\dots$$

$$\text{As } X > 0 \quad E[|X|] = E(X) = \sum_{n=1}^{\infty} 2^n \cdot \frac{1}{2^n} = \sum_{n=1}^{\infty} 1$$

$$\therefore E|X| = \infty$$

i.e. $E|X|$ diverges