

MTH9831 Homework 2 Theoretical Questions

Team 5

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4.7. **Gaussian conditioning.** Consider the Gaussian process (X_1, X_2) of mean 0 and covariance

$$C = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}.$$

- Find IID standard Gaussians (Z_1, Z_2) that are linear combinations of (X_1, X_2) .
- Write down (X_1, X_2) in terms of (Z_1, Z_2) .
- Compute $E[X_2|X_1]$.
- Compute $E[e^{aX_2}|X_1]$ for $a \in \mathbb{R}$. What is the conditional distribution of X_2 given X_1 ?

a). b). Assume $\begin{pmatrix} X_1 \\ X_2 \end{pmatrix} = M \begin{pmatrix} Z_1 \\ Z_2 \end{pmatrix}$. then $MM^T = C$.

$$M = C^{-1} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^T = \begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix}$$

$$\begin{cases} X_1 = Z_1 \\ X_2 = -Z_1 + Z_2 \end{cases} \Rightarrow \begin{cases} Z_1 = X_1 \\ Z_2 = X_1 + X_2 \end{cases}$$

$$c). E[X_2|X_1] = \frac{E[X_2 X_1]}{E[X_1^2]} X_1 = \frac{-1}{1} X_1 = -X_1.$$

$$d). E[e^{aX_2}|X_1] = E[e^{aZ_2 - aZ_1}|Z_1] \quad (\text{substitution})$$

$$= e^{-aZ_1} E[e^{aZ_2}|Z_1] \quad (\text{property of expectation}).$$

$$= e^{-aZ_1} E[e^{aZ_2}] \quad (Z_1, Z_2 \text{ are independent})$$

Since the MGF for $X \sim N(\mu, \sigma^2)$ is $M(t) = E[e^{tX}] = e^{t\mu + \frac{1}{2}\sigma^2 t^2}$

$$\text{We have } E[e^{aZ_2}] = e^{0 \cdot a + \frac{1}{2} \cdot 1 \cdot a^2} = e^{\frac{1}{2}a^2}.$$

$$E[e^{aX_2}|X_1] = e^{-aZ_1 + \frac{1}{2}a^2} = e^{-aX_1 + \frac{1}{2}a^2}$$

observe that $E[e^{aX_2}|X_1] = e^{(-X_1) \cdot a + \frac{1}{2} \cdot 1 \cdot a^2}$ is the MGF for

$N(-X_1, 1)$. Thus the conditional distribution of X_2

given X_1 follows $N(-X_1, 1)$. $f_{X_2|X_1}(x_2|x_1) = \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{1}{2}(x_2+x_1)^2}$

4.11. **Geometric Poisson process.** Let $(N_t, t \geq 0)$ be a Poisson process of intensity λ . For $\alpha > 0$, prove that the process $(e^{\alpha N_t - \lambda t(e^\alpha - 1)}, t \geq 0)$ is a martingale for the filtration of the Poisson process $(N_t, t \geq 0)$.

① Given α , $(e^{\alpha N_t - \lambda t(e^\alpha - 1)}, t \geq 0)$ is adapted to the filtration of $(N_t, t \geq 0)$

$$\begin{aligned} \textcircled{2} \quad E[e^{\alpha N_t - \lambda t(e^\alpha - 1)}] &= E[e^{\alpha N_t - \lambda t(e^\alpha - 1)}] \\ &= \sum_{n=0}^{\infty} \frac{(\lambda t)^n}{n!} e^{-\lambda t} \cdot e^{\alpha n - \lambda t(e^\alpha - 1)} \\ &= \sum_{n=0}^{\infty} \frac{(\lambda t)^n}{n!} e^{\alpha n - \lambda t e^\alpha} \\ &= \sum_{n=0}^{\infty} \frac{(\lambda t e^\alpha)^n}{n!} \cdot e^{-\lambda t e^\alpha} = e^{\lambda t e^\alpha} \cdot e^{-\lambda t e^\alpha} = 1 \end{aligned}$$

$$\begin{aligned} \textcircled{3} \quad E[e^{\alpha N_t - \lambda t(e^\alpha - 1)} | N_s] &= E[e^{\alpha N_s - \lambda s(e^\alpha - 1)} \cdot e^{\alpha(N_t - N_s) - \lambda(e^\alpha - 1)(t-s)} | N_s] \\ &= e^{\alpha N_s - \lambda s(e^\alpha - 1)} \cdot E[e^{\alpha(N_t - N_s) - \lambda(e^\alpha - 1)(t-s)}] \\ &= e^{\alpha N_s - \lambda s(e^\alpha - 1)} \cdot E[e^{\lambda(t-s)(e^\alpha - 1)} e^{-\lambda(e^\alpha - 1)(t-s)}] \\ &= e^{\alpha N_s - \lambda s(e^\alpha - 1)} \end{aligned}$$

Thus the process $(e^{\alpha N_t - \lambda t(e^\alpha - 1)}, t \geq 0)$ is a martingale for the filtration of the Poisson process $(N_t, t \geq 0)$

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Q12 a) we have $M_t = tB_t - \frac{1}{3}B_t^3$

① Adaptedness: M_t consists of B_t which is an \mathcal{F}_t -adapted process. Hence M_t is also \mathcal{F}_0 -adapted.

② Integrable: $E[|M_{t+1}|] \leq E[|tB_t|] + \frac{1}{3}E[|B_t^3|]$
 $\leq |t| \cdot E[B_t^2]^{1/2} + \frac{1}{3}E[B_t^4]^{1/2}$

(use that $t \geq 0$) $= t^{3/2} + 5t^{3/2} = 6t^{3/2} < \infty$, for any given $t > 0$.

Hence M_t is integrable.

③ Martingale property. $\forall t > s, E[M_t | \mathcal{F}_s] = E[tB_t - \frac{1}{3}B_t^3 | \mathcal{F}_s]$
 $= E[tB_t | \mathcal{F}_s] - \frac{1}{3}E[B_t^3 | \mathcal{F}_s]$

$E[tB_t | \mathcal{F}_s] = tB_s$ since BM is a martingale, t is a constant

$E[B_t^3 | \mathcal{F}_s] = E[(B_t - B_s + B_s)^3 | \mathcal{F}_s] = E[(B_t - B_s)^3 | \mathcal{F}_s] + 3E[(B_t - B_s)^2 B_s | \mathcal{F}_s] + 3E[(B_t - B_s) B_s^2 | \mathcal{F}_s] + E[B_s^3 | \mathcal{F}_s]$

since $(B_t - B_s)$ is independent of B_s (hence \mathcal{F}_s), we have

$E[(B_t - B_s)^3 | \mathcal{F}_s] = E[(B_t - B_s)^3] = 0$ (third moment of $N(0, t-s)$)

$E[(B_t - B_s)^2 B_s | \mathcal{F}_s] = B_s E[(B_t - B_s)^2] = (t-s)B_s$; $E[(B_t - B_s) B_s^2 | \mathcal{F}_s] = B_s^2 E[B_t - B_s] = 0$

$E[B_s^3 | \mathcal{F}_s] = B_s^3$, where we also use that B_s is \mathcal{F}_s -measurable. Putting together:

$E[M_t | \mathcal{F}_s] = tB_s - \frac{1}{3} \cdot 3(t-s)B_s - \frac{1}{3}B_s^3 = sB_s - \frac{1}{3}B_s^3 = M_s$.

Hence, $E[M_t | \mathcal{F}_s] = M_s$ for $s < t$.

By ①, ②, ③, we've shown that $(M_t, t \geq 0)$ is a martingale. □

b) we first show that $T < \infty$ a.s.

$P(T < \infty) \geq P(\bigcup_{n=1}^{\infty} \{ |B_{n+1} - B_n| > a+b \}) = 1 - P(\bigcap_{n=1}^{\infty} \{ |B_{n+1} - B_n| \leq a+b \}) = 1 - \lim_{m \rightarrow \infty} P(\bigcap_{n=1}^m \{ |B_{n+1} - B_n| \leq a+b \})$
 (where $p = P(|B_{i+1} - B_i| \leq a+b) < 1$) continuity of probability
 $\leq 1 - \lim_{m \rightarrow \infty} p^m = 1$.

Hence $T < \infty$ a.s.

Note that, $M_{t \wedge T}$ is bounded since $B_{t \wedge T} \in [-b, a]$ and $T < \infty$ a.s. Doob's Optional Stopping theorem: we can hence apply

theorem: $E[M_T] = E[M_0]$

$\Leftrightarrow E[M_T] = 0$ since $B_0 = 0$

$\Leftrightarrow E[\tau B_T] = \frac{1}{3}E[B_T^3] = \frac{1}{3} \left(\frac{ba^3}{a+b} - \frac{ab^3}{a+b} \right) = \frac{1}{3} \frac{ab}{a+b} \cdot (a-b)(a+b) = \frac{1}{3} ab(a-b)$.

since from the lecture, we know that $P(B_T = a) = \frac{b}{a+b}$ and $P(B_T = -b) = \frac{a}{a+b}$. □

c) we have $E[e^{aB_T - \frac{a^2}{2}T}] = E[E[e^{aB_T - \frac{a^2}{2}T} | \mathcal{F}_t]] \dots (*)$

Conditioned on $T=t$, we have:

$\mathbb{E}[e^{aB_t - \frac{1}{2}a^2 t}] = e^{-\frac{1}{2}a^2 t} \mathbb{E}[e^{aB_t}] = e^{-\frac{1}{2}a^2 t} \mathbb{E}[e^{aX}]$ for $X \sim N(0, t)$
 $\mathbb{E}[e^{aX}]$ is just the m.g.f. for $N(0, t)$ valued at a , hence $\mathbb{E}[e^{aX}] = e^{\frac{1}{2}ta^2}$.
 Thus, $e^{-\frac{1}{2}a^2 t} \cdot \mathbb{E}[e^{aB_t}] = e^{-\frac{1}{2}a^2 t} e^{\frac{1}{2}a^2 t} = 1, \forall a > 0$.
 Thus, (*) becomes $\mathbb{E}[1] = 1$, and hence $\mathbb{E}[e^{aB_t - \frac{1}{2}a^2 t}] = 1 \quad \forall a > 0$. □

d) By c), we can write $\exp(aB_t - \frac{1}{2}a^2 t)$ as a Taylor series around 0:
 $\mathbb{E}[e^{aB_t - \frac{1}{2}a^2 t}] = 1 \Leftrightarrow 1 = \mathbb{E}[1 + aB_t - \frac{1}{2}a^2 t + \frac{1}{2}(aB_t - \frac{1}{2}a^2 t)^2 + \frac{1}{6}(aB_t - \frac{1}{2}a^2 t)^3 + \dots]$
 $1 = 1 + a\mathbb{E}[B_t] + \frac{1}{2}a^2 \mathbb{E}[B_t^2 - t] - \frac{1}{6}a^3 \mathbb{E}[3tB_t] + \frac{1}{6}a^3 \mathbb{E}[B_t^3] + \sum_{n \geq 4} f_n a^n$
 $\Leftrightarrow 0 = a\mathbb{E}[B_t] + \frac{1}{2}a^2 \mathbb{E}[B_t^2 - t] + \frac{1}{6}a^3 \mathbb{E}[B_t^3 - 3tB_t] + \sum_{n \geq 4} f_n a^n$, where f_n is a function of B_t and t .
 LHS = 0 \Leftrightarrow all coefficients = 0. Since B_t and $B_t^2 - t$ are martingales (shown in class)
 $\Rightarrow \mathbb{E}[B_t] = \mathbb{E}[B_0] = 0, \mathbb{E}[B_t^2 - t] = \mathbb{E}[B_0^2] = 0$. Then we must also have $\mathbb{E}[B_t^3 - 3tB_t] = 0$, which means
 $\mathbb{E}[3tB_t] = \frac{1}{3} \mathbb{E}[B_t^3] = \frac{1}{3} \left(\frac{b_0^3}{a+b} - \frac{ab^3}{a+b} \right) = \frac{1}{3} ab(a-b)$.
 this is the required result. □

Q15. a) $\tau: \min_{t \geq 0} \{X_t = +1\}$. So this means: Stop as soon as winning a game.

Now, $\mathbb{E}[M_\tau] = \mathbb{E}[\sum_{n \geq 1} \mathbb{1}_{\{\tau \geq n\}} M_n] = \sum_{n \geq 1} \mathbb{E}[\mathbb{1}_{\{\tau \geq n\}} M_n]$ by Fubini's thm (Ex. 3.11)
 when $\tau = n$, we have: $M_\tau = 1 \cdot (-1) + 2 \cdot (-1) + \dots + 2^{n-1} \cdot (-1) + 2^n \cdot 1 = 2^n - (1 + 2 + \dots + 2^{n-1}) = 1$.
 so we can rewrite it as $\sum_{n \geq 1} \mathbb{E}[\mathbb{1}_{\{\tau \geq n\}} \cdot 1] = \sum_{n \geq 1} \mathbb{E}[\mathbb{1}_{\{\tau \geq n\}}] = \sum_{n \geq 1} \mathbb{P}(\tau \geq n)$
 $= \sum_{n \geq 1} (\frac{1}{2})^n$.

since $\sum_{n \geq 1} (\frac{1}{2})^n = \frac{1}{2} \cdot (\sum_{n \geq 0} (\frac{1}{2})^n) = \frac{1}{2} \cdot \frac{1}{1 - \frac{1}{2}} = 1$, we have $\mathbb{E}[M_\tau] = 1$.

What about $\mathbb{E}[M_0]$? $\mathbb{E}[M_0] = \mathbb{E}[S_0] = 0$. This proves the required result. □

b) we can't apply Optional Stopping since $M_{n \wedge \tau}$ is not bounded.

$\forall m \in \mathbb{N}, m > 0$, we have when $n > \tau$, $M_{n \wedge \tau} = M_n = -1 - 2 - \dots - 2^{n-1} = 1 - 2^{n+1}$
 $\Rightarrow M_{n \wedge \tau} = 1 - 2^{n+1} < -m$ if $n > \log_2 m + 1$. Thus $M_{n \wedge \tau}$ is not bounded below.
 \Rightarrow violation of Doob's Optional Stopping theorem assumptions. □

c) Disadvantages:

- ① Takes a lot to win (Input might be infinity) A player might lose all his/her money before winning.
- ② Even if wins, player only wins 1 (little return).

Q16 proof

- ① Adaptedness. $M_t = \mathbb{E}[X | \mathcal{F}_t]$ which is \mathcal{F}_t -measurable for any $t \geq 0$
 Hence M_t is an \mathcal{F}_t -adapted process.
- ② Integrable. we have that $\mathbb{E}[M_t] = \mathbb{E}[\mathbb{E}[X | \mathcal{F}_t]]$.

By Triangle Inequality (Jensen's Inequality on 1.1), $|\mathbb{E}[X|\mathcal{F}_t]| \leq \mathbb{E}[|X||\mathcal{F}_t]$. Thus, $\forall t \geq 0$
 $\mathbb{E}[|M_t|] \leq \mathbb{E}[\mathbb{E}[|X||\mathcal{F}_t]] = \mathbb{E}[|X|] < \infty$ as $|X|$ is integrable.

Hence M_t is integrable.

③ Martingale property, $\forall t > s > 0$, we have: $\mathbb{E}[M_t|\mathcal{F}_s] = \mathbb{E}[\mathbb{E}[X|\mathcal{F}_t]|\mathcal{F}_s]$.

since $(\mathcal{F}_t, t \geq 0)$ is a filtration and $s < t$, we have that $\mathcal{F}_s \subseteq \mathcal{F}_t$. By the tower property, we have: $\mathbb{E}[M_t|\mathcal{F}_s] = \mathbb{E}[\mathbb{E}[X|\mathcal{F}_t]|\mathcal{F}_s] = \mathbb{E}[X|\mathcal{F}_s] = M_s$.

By ①, ②, ③, we conclude that $M_t = \mathbb{E}[X|\mathcal{F}_t]$ is a martingale.

□