

Problem Set 8: Stochastic Calculus

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Problem A-1.

- (a) True.
- (b) False.
- (c) True. $\mathbb{E}[X_{t+1}|X_t] = X_t + \mathbb{E}[Y_t - \frac{1}{\lambda}] = X_t$
- (d) False. $\mathbb{E}[X_{t+1}|X_t] = X_t + \mathbb{E}[Z_t Z_{t-1} \dots Z_0] = X_t + \mathbb{E}[e^{W_t}] = X_t + (t+1)\mu + \frac{t+1}{2}\sigma^2$, where $W_t \sim \mathcal{N}((t+1)\mu, (t+1)\sigma^2)$
- (e) True. $\mathbb{E}[X_{t+1}|X_t] = \mathbb{E}[B_{t+1}^{(1)} B_{t+1}^{(2)} | B_t^{(1)} B_t^{(2)}] = \mathbb{E}[(B_t^{(1)} + Z_1)(B_t^{(2)} + Z_2)] = B_t^{(1)} B_t^{(2)}$, where $Z_1 \sim \mathcal{N}(0, 1)$, $Z_2 \sim \mathcal{N}(0, 1)$, and $Z_1 \perp Z_2$.

Problem A-2.

- (a) $\mathbb{E}[B(t)|B(s)] = B(s)$ (Martingale). $\text{Var}[B(t)|B(s)] = \text{Var}[B(s) + Z] = t - s$.
- (b) $\mathbb{E}[X(t)|X(s)] = \mu(t - s) + B(s)$. $\text{Var}[X(t)|X(s)] = t - s$.
- (c) $\mathbb{E}[e^{\sigma(B(t)-B(s))}] = \mathbb{E}[e^{\sigma Z}] = \frac{t-s}{2}\sigma^2$, where $Z \sim \mathcal{N}(0, t - s)$.

Problem A-3.

- (a) False. Y_t depends on the future of X_t .
- (b) False. $Y_t = \max_{0 \leq s \leq 2t} X_s$ also depends on the future of X_t .
- (c) True.

Problem A-4.

- (a) $df = 3B_t dt + 3B_t^2 dB_t$
- (b) $df = -\frac{1}{2} \sin B_t dt + \cos B_t dB_t$
- (c) $df = -[(1 + 3t^2) \sin(t^2 + B_t^2) + 2 \cos(t^3 + B_t^2) B_t^2] dt - 2 \cos(t^3 + B_t^2) B_t dB_t$
- (d) $df = (1 + 2B_t^2) e^{B_t^2} dt + 2B_t e^{B_t^2} dB_t$
- (e) $df = B_t^2 dB_t$
- (f) $df = B_t dt$
- (g) $df = (2\mu X_t + \sigma^2) dt + 2\sigma X_t dB_t$

Problem A-5.

Proof: Let $p(x)$ be the pdf given by B_t , $\tilde{p}(x)$ be the pdf given by B_t^2 . For all $X \subseteq (-\infty, 0)$, we have $p(X) > 0$ while $\tilde{p}(X) = 0$. Therefore, $B(t)$ and $B(t)^2$ are not equivalent.

Problem B-1.

(i) For a fixed value t , we have $X_1(t) = Z_1 \sim \mathcal{N}(0, t)$ and $X_2(t) = Z_2 \sim \mathcal{N}(0, t)$, and $Z_1 \perp Z_2$.

$$p_Y(z_1, z_2) = p_{Z_1}(z_1)p_{Z_2}(z_2) = \frac{1}{2\pi t} e^{-\frac{z_1^2 + z_2^2}{2t}} \quad (8.1)$$

$$(ii) P(Y(t) \in D_\rho) = \int_{-\pi}^{\pi} \int_0^\rho r \cdot \frac{1}{2\pi t} e^{-\frac{r^2}{2t}} dr d\theta = \int_0^\rho \frac{r}{t} e^{-\frac{r^2}{2t}} dr = 1 - e^{-\frac{\rho^2}{2t}}$$

Problem B-2.

Let $\Delta(t) = B_t$, which is definitely adapted to B_t , then by using Ito isometry, we have

$$\mathbb{E}[(\int_0^t B_s dB_s)^2] = \mathbb{E}[\int_0^t B_s^2 ds] = \int_0^t \mathbb{E}[B_s^2] ds = \frac{1}{2} t^2 \quad (8.2)$$

Problem B-3.

$$(i) \int_0^t h(s) dB_s = h(s) B_s|_0^t - \int_0^t B_s h'(s) ds = h(t) B_t - \int_0^t B_s h'(s) ds$$

(ii) Let $h(t) = t$. By using part (i), we have

$$\int_0^T s dB_s = T B_T - \int_0^T B_s ds. \quad (8.3)$$

Then we can obtain $\int_0^T B_s ds = T B_T - \int_0^T s dB_s = \int_0^T T dB_s - \int_0^T s dB_s = \int_0^T (T-s) dB_s$, which can be seen as the sum of independent normal random variables. Therefore, $\int_0^T B_s ds$ is also normal distributed.