

Stochastic Differential Equations

Homework Sheet 3 - solutions

Problem 1. Show that

$$\frac{1}{(2\pi)^{d/2}} \int_{\mathbb{R}^d} dz \{ |z|^2 \}^2 e^{-\frac{1}{2}|z|^2} = d(d+2), \quad (1)$$

where $|z|^2 := \sum_{i=1}^d (z_i)^2$. Conclude that for $\mathbb{R}$ -valued random variable $Z \sim N(0, \sigma^2)$ we have $E(Z^4) = 3\sigma^4$.

Solution. By Problem 5 of HS2

$$I(a) := \int_{\mathbb{R}^d} e^{-\frac{a}{2}|z|^2} dz = (2\pi)^{d/2} a^{-d/2}, \quad a > 0.$$

Differentiating under the integral,

$$\begin{aligned} I'(a) &= \int_{\mathbb{R}^d} \left(-\frac{1}{2}|z|^2\right) e^{-\frac{a}{2}|z|^2} dz, \\ I''(a) &= \int_{\mathbb{R}^d} \left(-\frac{1}{2}|z|^2\right)^2 e^{-\frac{a}{2}|z|^2} dz. \end{aligned}$$

Hence

$$\int_{\mathbb{R}^d} (|z|^2)^2 e^{-\frac{a}{2}|z|^2} dz = 4 I''(a).$$

From the explicit formula for $I(a)$ above

$$I''(a) = (2\pi)^{d/2} \frac{d}{2} \left(\frac{d}{2} + 1 \right) a^{-\frac{d}{2}-2}.$$

Therefore

$$\int_{\mathbb{R}^d} (|z|^2)^2 e^{-\frac{a}{2}|z|^2} dz = (2\pi)^{d/2} d(d+2) a^{-\frac{d}{2}-2}.$$

Thus for $a = 1$ we obtain the claim.

Now if $d = 1$ and $Z \in N(0, \sigma^2)$, then $Z/\sigma \in N(0, 1)$ (see lecture notes, construction of Brownian motion via the CLT). Then

$$E[(Z/\sigma)^4] = d(d+2) = 3, \quad (2)$$

hence, $E(Z^4) = 3\sigma^4$.

Problem 2. Let $Y \sim N(0, 1)$ and set for some $a > 0$

$$Z = Y\chi_{\{|Y| \leq a\}} - Y\chi_{\{|Y| > a\}}. \quad (3)$$

Show that also $Z \sim N(0, 1)$. Also, show that $Y + Z$ is not normal, hence $X = (Y, Z)$ is not a multi-normal random variable. (In this exercise non-degenerate (multi-)normal distributions are meant).

Notation: $\chi_{\{|Y| \leq a\}}(\omega) := \chi_{\{y \in \mathbb{R} : |y| \leq a\}}(Y(\omega))$. The function $\chi_{\{|Y| > a\}}$ is defined analogously.

Solution. Step 1: Z is standard normal.

Let $\phi_Z(u) = E[e^{iuZ}]$. Then, we verify that, for any characteristic function χ_A ,

$$e^{iuY\chi_A} = \chi_{A^c} + e^{iuY}\chi_A, \quad (4)$$

by checking separately for $\omega \in A$ and $\omega \in A^c$. Setting $A = \{|Y| \leq a\}$ and $A^c = \{|Y| > a\}$, we have

$$e^{iuZ} = e^{iuY\chi_A} e^{-iuY\chi_{A^c}} = (\chi_{A^c} + e^{iuY}\chi_A)(\chi_A + e^{-iuY}\chi_{A^c}) = e^{iuY}\chi_A + e^{-iuY}\chi_{A^c}. \quad (5)$$

Hence,

$$\phi_Z(u) = E[e^{iuY}\chi_{\{|Y| \leq a\}}] + E[e^{-iuY}\chi_{\{|Y| > a\}}].$$

Compute the second term explicitly via the density of Y :

$$E[e^{-iuY}\chi_{\{|Y| > a\}}] = \int_{|y| > a} e^{-iuy} \frac{e^{-y^2/2}}{\sqrt{2\pi}} dy.$$

With the change of variables $x = -y$ we get

$$\int_{|y| > a} e^{-iuy} \frac{e^{-y^2/2}}{\sqrt{2\pi}} dy = \int_{|x| > a} e^{iux} \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx = E[e^{iuY}\chi_{\{|Y| > a\}}].$$

Hence

$$\phi_Z(u) = E[e^{iuY}\chi_{\{|Y| \leq a\}}] + E[e^{iuY}\chi_{\{|Y| > a\}}] = E[e^{iuY}] = e^{-u^2/2}.$$

Therefore $Z \sim N(0, 1)$.

Step 2: $Y + Z$ is not normal. By definition,

$$Y + Z = \begin{cases} 2Y, & |Y| \leq a, \\ 0, & |Y| > a. \end{cases}$$

If $|Y| \leq a$ then $Y \leq a$ so $2Y \leq 2a$, hence $(Y + Z) > 2a$ cannot occur in this case. If $|Y| > a$ then $Y + Z = 0 \leq 2a$, so $(Y + Z) > 2a$ also cannot occur. Therefore

$$P(Y + Z > 2a) = 0.$$

But any non-degenerate Gaussian random variable has an everywhere strictly positive density. Consequently, $Y + Z$ cannot be such a Gaussian.

Step 3: (Y, Z) is not jointly normal. If (Y, Z) were jointly normal (non-degenerate), then every nontrivial linear combination, in particular $Y + Z$, would be normal (non-degenerate), contradicting Step 2. Thus $X = (Y, Z)$ is not multi-normal.

Problem 3. Suppose $X_k : \Omega \rightarrow \mathbb{R}^d$ are normal for all k and $X_k \rightarrow X$ in $L^2(\Omega)$, i.e.

$$E[|X_k - X|^2] \rightarrow 0 \quad \text{as} \quad k \rightarrow \infty. \quad (6)$$

Show that X is normal.

Solution. The most direct solution appear to be to write

$$E[\exp(i\langle u, X_k \rangle)] = \exp\left(-\frac{1}{2}\langle u, C_k u \rangle + i\langle u, m_k \rangle\right) \quad (7)$$

and then use the formulas from the lecture expressing C_k and m_k in terms of X_k (as done in class).

Here we give a different argument: First, note that $|e^{i\langle u, x \rangle} - e^{i\langle u, y \rangle}| \leq |u| \cdot |x - y|$. This follows, e.g., from

$$\begin{aligned} |e^{i\langle u, x \rangle} - e^{i\langle u, y \rangle}| &= |1 - e^{i\langle u, (x-y) \rangle}| \\ &= \left| \int_0^1 \frac{d}{dw} e^{iw\langle u, (x-y) \rangle} dw \right| \\ &= |i\langle u, (x-y) \rangle \int_0^1 e^{iw\langle u, (x-y) \rangle} dw| \\ &\leq |u||x-y|. \end{aligned} \quad (8)$$

Thus, we have

$$E\left[\left\{\exp(i\langle u, X_k \rangle) - \exp(i\langle u, X \rangle)\right\}^2\right] \leq |u|^2 E[|X_k - X|^2] \quad (9)$$

as $k \rightarrow \infty$. Therefore

$$\begin{aligned} |E[\exp(i\langle u, X_k \rangle)] - E[\exp(i\langle u, X \rangle)]| &\leq \int |\exp(i\langle u, X_k \rangle) - \exp(i\langle u, X \rangle)| dP \\ &\leq \left(\int |\exp(i\langle u, X_k \rangle) - \exp(i\langle u, X \rangle)|^2 dP\right)^{1/2} \rightarrow 0, \end{aligned} \quad (10)$$

(11)

where we used the Cauchy-Schwarz inequality. Up to now we have not used that X_k are Gaussian. Actually, we have verified, quite generally, that the L^2 -convergence implies the convergence of the characteristic function, pointwise in u .

We conclude that

$$E[\exp(i\langle u, X_k \rangle)] = \exp\left(-\frac{1}{2}\langle u, C_k u \rangle + i\langle u, m_k \rangle\right) \quad (12)$$

converge for any fixed u .

Using this, let us now justify that C_k and m_k converge to some C and m . In fact, since (12) converges, also its absolute value does, hence $\exp(-\frac{1}{2}\langle u, C_k u \rangle)$ must converge, therefore also $\langle u, C_k u \rangle$ must converge. As this holds for all u , by polarization,

$$\langle v_1, C_k v_2 \rangle = \frac{1}{4}(\langle (v_1 + v_2), C_k (v_1 + v_2) \rangle - \langle (v_1 - v_2), C_k (v_1 - v_2) \rangle) \quad (13)$$

all matrix elements converge. Now it suffices to show that convergence of $e^{i u m_k}$ for all u implies convergence of m_k . We proceed by contradiction. First, suppose $m_k \rightarrow \infty$ (perhaps along a subsequence). Set $F(u) := \lim_{k \rightarrow \infty} e^{i u m_k}$. Then, by the Riemann-Lebesgue lemma and dominated convergence, for any $f \in L^1(\mathbb{R}^d)$

$$0 = \lim_{k \rightarrow \infty} \int e^{i u m_k} f(u) du = \int F(u) f(u) du. \quad (14)$$

But F is non-zero since $F(0) = 1$ and we can choose e.g. $f(u) = \overline{F}(u) e^{-|u|^2}$ to ensure that the integral (14) is non-zero and finite, which is a contradiction. Now suppose that m_k has two (or more) finite accumulation points. Choose any $g \in L^1(\mathbb{R}^d)$ s.t. $\hat{g}$ takes different values near these accumulation points. Then

$$\int e^{i u m_k} g(u) du = \hat{g}(m_k). \quad (15)$$

By taking the limit of both sides the l.h.s. converges by dominated convergence and the r.h.s. does not as $k \mapsto \hat{g}(m_k)$ jumps between the two accumulation points. Again, a contradiction.

Since $C_k = C_k^T$ for all k , also the limiting matrix is symmetric. Similarly, since for any $u \in \mathbb{R}^d$ we have $\langle u, C_k u \rangle \geq 0$ and the inequality $\geq$ survives limits, we conclude that C has positive eigenvalues.

But we cannot conclude that a sequence of non-degenerate Gaussians converges to a non-degenerate Gaussian, since the inequality $>$ may become $\geq$ in the limit. Actually, such a scenario can easily happen under the assumptions of this problem: Let Z be any non-degenerate Gaussian, set $X_k = \frac{1}{k} Z$ so that it trivially converges in $L^2(\Omega)$ to $X \equiv 0$. Note that $X \equiv 0$ has the Dirac delta distribution hence it is degenerate Gaussian.

Problem 4. Let $\{X_n\}_{n \geq 1}$ and X be random variables on a common probability space. We say that X_n converges in probability to X , written $X_n \xrightarrow{P} X$, if for every $\varepsilon > 0$,

$$\lim_{n \rightarrow \infty} P(|X_n - X| > \varepsilon) = 0.$$

Show that if $X_n \rightarrow X$ in $L^2(\Omega)$, i.e.

$$E[(X_n - X)^2] \xrightarrow{n \rightarrow \infty} 0,$$

then $X_n \xrightarrow{P} X$. Hint: Use the Chebyshev's inequality from HS2, Problem 3.

Solution. Will be discussed in class on 07.11.2025 and posted afterwards.

Problem 5. Let $\{B_t\}_{t \in \mathbb{R}_+}$ be d -dimensional Brownian motion, $d \geq 3$, starting at x . Let $K \subset \mathbb{R}^d$ be a bounded Borel set and $T > 0$. Then the random variable

$$\Omega \ni \omega \mapsto \int_0^T \chi_{\{\tilde{\omega} \in \Omega: B_t(\tilde{\omega}) \in K\}}(\omega) dt \quad (16)$$

denoted in short-hand notation

$$\int_0^T \chi_{\{B_t \in K\}} dt = \lim_{|\Pi| \rightarrow 0} \sum_{i=1}^n \chi_{\{B_{\xi_i} \in K\}} \Delta t_i \quad (17)$$

is the time this stochastic process spends in K up to the ‘time horizon’ T . (This is more clear from the Riemann sum representation, where the sum counts only the time-intervals Δt_i for which the Brownian motion B_{ξ_i} , $\xi_i \in [t_{i-1}, t_i]$, is in K). Show that

$$\lim_{T \rightarrow \infty} E^x \left[\int_0^T \chi_{\{B_t \in K\}} dt \right] = \int_K G(x, y) dy, \quad (18)$$

where $G(x, y) := \frac{c_d}{|x-y|^{d-2}}$ for some $c_d > 0$.

Solution. Will be discussed in class on 07.11.2025 and posted afterwards.

To be discussed in class: 24.10.2025