

Stochastic analysis, autumn 2011, Exercises-2, 23.09.11

1. Let

$$f(x) = f(0) + \int_0^x \dot{f}(y)dy, \quad h(x) = h(0) + \int_0^x \dot{h}(y)dy,$$

absolutely continuous function with $\dot{f}, \dot{h} \in L^2(\mathbb{R}, \gamma)$ where $\gamma(dx) = \phi(x)dx$ denotes the standard gaussian measure.

- For a standard gaussian random variable $G(\omega)$ with $E(G) = 0$, $E(G) = 1$ prove the gaussian integration by parts formula:

$$E_P \left(f'(G)h(G) \right) = E_P \left(f(G)(Gh(G) - h'(G)) \right)$$

Hint: rewrite the expectation as integral, and use integration by parts with respect to Lebesgue measure.

We show first that $E(f'(G)^2) < \infty$ implies $E(f(G)^2) < \infty$.

$$\begin{aligned} & E \left((f(G) - f(0))^2 \mathbf{1}(G \geq 0) \right) \int_0^\infty \left(\int_0^x f'(y)dy \right)^2 \phi(x)dx \\ & \leq \int_0^\infty \int_0^x f'(y)^2 \phi(x)dydx \\ & = \int_0^\infty \left(\int_y^\infty \phi(x)dx \right) f'(y)^2 dy \leq \int_0^\infty f'(y)^2 \min(1, \frac{1}{y}) \phi(y)dy \\ & \leq E(f'(G)^2 \mathbf{1}(G \geq 0)) \end{aligned}$$

where we have used Cauchy-Schwartz, Fubini, and a gaussian tail bound. Similarly

$$E((f(G) - f(0))^2 \mathbf{1}(G \leq 0)) \leq E(f'(G)^2 \mathbf{1}(G \geq 0))$$

which implies by the triangle inequality

$$E(f(G)^2)^{1/2} \leq E(f'(G)^2)^{1/2} + |f(0)| < \infty$$

Therefore it is enough to assume $f'(G), h'(G) \in L^2(\mathbb{R}, \gamma)$.

Integrating by parts,

$$\begin{aligned} & f(b)h(b)\phi(b) - f(a)h(a)\phi(a) = \\ & \int_a^b f'(x)h(x)\phi(x)ds + \int_a^b f(x)h(x)'\phi(x)ds + \int_a^b f(x)h(x)\phi(x)'\phi(x)ds \\ & = \int_a^b f'(x)h(x)\phi(x)ds + \int_a^b f(x)h(x)'\phi(x)ds - \int_a^b x f(x)h(x)\phi(x)ds \end{aligned}$$

since for the standard gaussian density $\phi'(x) = -x\phi(x)$. We show that $f(x)h(x)\phi(x) \rightarrow 0$ as $x \rightarrow \pm\infty$. Since

$$\int_0^\infty |f(x)h(x)|\phi(x)dx \leq \sqrt{E(f(G)^2)E(h(G)^2)} < \infty$$

we have

$$\lim_{b \rightarrow \infty} \int_b^{\infty} |f(x)h(x)|\phi(x)dx = 0$$

and since $f(b)h(b)$ is an absolutely continuous function with integrable derivative

$$E(|f'(G)h(G)| + f(G)h'(G)) \leq E(|f'(G)h(G)|) + E(|h'(G)h(G)|) < \infty$$

necessarily

$$\lim_{b \rightarrow \infty} f(b)h(b)\phi(b) = 0$$

otherwise there is $x_n \rightarrow \infty$ such that $|f(x_n)h(x_n)\phi(x_n)| > \varepsilon$. But $\forall \eta > 0$ and $\forall n$ large enough, $r > 0$

$$\phi(x_n + r) \int_{x_n}^{x_n+r} \left| \frac{d}{dy}(fh)(y) \right| dy \leq \int_{x_n}^{x_n+r} \left| \frac{d}{dy}(fh)(y) \right| \phi(y) dy \leq \int_{x_n}^{\infty} \left| \frac{d}{dy}(fh)(y) \right| \phi(y) dy < \eta$$

which gives

$$|f(x_n + r)h(x_n + r)| > \frac{\varepsilon}{\phi(x_n)} - \frac{\eta}{\phi(x_n + r)}.$$

This gives

$$\begin{aligned} \int_{x_n}^{x_n+\Delta} |f(x)h(x)|\phi(x)dx &> \int_0^{\Delta} \left(\varepsilon \frac{\phi(y)}{\phi(x_n)} - \eta \right) dy \\ &= \varepsilon \int_0^{\Delta} \exp\left(-\frac{r^2}{2} + 2rx_n\right) dr - \eta\Delta \rightarrow \infty \text{ as } x_n \rightarrow \infty \end{aligned}$$

for fixed $\eta > 0$ and $\Delta > 0$ This is in contradiction with $E(|f(G)h(G)|) < \infty$.

By taking limits as $a \rightarrow -\infty$, $b \rightarrow +\infty$ we obtain

$$\begin{aligned} 0 &= \int_{\mathbb{R}} f'(x)h(x)\phi(x)ds + \int_{\mathbb{R}} f(x)h(x)'\phi(x)ds - \int_{\mathbb{R}} xf(x)h(x)\phi(x)ds = \\ &E(f'(G)h(G)) + E(f(G)(h'(G) - Gh(G))) \end{aligned}$$

- Write the corresponding gaussian integration by parts for $B_T(\omega) \sim \mathcal{N}(0, T)$ with $T > 0$

Solution Take $B_T(\omega) = \sqrt{T}G(\omega)$ with $G(\omega) \sim \mathcal{N}(0, 1)$

Then

$$\begin{aligned} E_P(f'(B_T)h(B_T)) &= E_P(f'(\sqrt{T}G)h(\sqrt{T}G)) = E_P\left(\frac{1}{\sqrt{T}} \frac{d}{dx} f(\sqrt{T}x) \Big|_{x=G} h(\sqrt{T}G)\right) \\ &= \frac{1}{\sqrt{T}} E_P\left(f(\sqrt{T}G) \left(h(\sqrt{T}G)G - \frac{d}{dx} h(\sqrt{T}x) \Big|_{x=G} \right)\right) \\ &= E_P\left(f(\sqrt{T}G) \left(\frac{h(\sqrt{T}G)\sqrt{T}G}{T} - h'(\sqrt{T}G) \right)\right) \\ &= E_P\left(f(B_T) \left(\frac{h(B_T)B_T}{T} - h'(B_T) \right)\right) \end{aligned}$$

2. Show that a process X_t with independent increments is Markov, which means for $0 \leq s \leq t$ and $\mathcal{F}_s^X = \sigma(X_u : 0 \leq u \leq s)$ and a bounded measurable test function $f(x)$

$$E(f(X_t)|\mathcal{F}_s^X) = E(f(X_t)|\sigma(X_s))$$

Hint. use first the decomposition $X_t = X_s + (X_t - X_s)$ and the definition of conditional expectation to show first that for a bounded measurable test-function

$$E(f(X_t)|\sigma(X_s))(\omega) = E\left(f(y + (X_t - X_s))\right)\Big|_{y=X_s(\omega)}$$

Solutions We will use the following lemma: if X and Y are random variables on (Ω, \mathcal{F}, P) , $\mathcal{G} \subseteq \mathcal{F}$ is a sub- σ -algebra and Y is \mathcal{G} -measurable, and X is P -independent from the σ -algebra \mathcal{G} , then for every non-negative Borel-measurable function $f(x, y)$, and X

$$E_P(f(X, Y)|\mathcal{G})(\omega) = E_P(f(X, y))\Big|_{y=Y(\omega)}$$

Proof: it is true for $f(x, y) = f_1(x)f_2(y)$, by using independence and the definition of conditional expectation. every non-negative Borel-measurable function $f(x, y)$ can be approximated from below by finite linear combinations of such product functions and the results follows by monotone convergence of conditional expectation.

Now for $0 \leq s \leq t$

$$\begin{aligned} E(f(X_t)|\mathcal{F}_s^X) &= E_P(f(X_s + (X_t - X_s))|\mathcal{F}_s^X) \\ &= E_P(f(y + X_t - X_s))\Big|_{y=X_s} = E_P(f(X_s + (X_t - X_s))|\sigma(X_s)) = E_P(f(X_t)|\sigma(X_s)) \end{aligned}$$

where we used the previous lemma twice.

3. $(N_t(\omega) : t \geq 0)$ is a λ -Poisson process ($\lambda > 0$) if $N_0 = 0$ and it has independent increments with $(N_t - N_s) \sim \text{Poisson}(\lambda(t - s))$

- Show that the Poisson process is non-decreasing with piecewise constant trajectories, it is P -almost surely finite on and increases only by jumps of size 1.

Hint: Show that

$$P\left(\exists t \leq T : \Delta N_t \geq 2\right) = 0$$

Write it as limits of probabilities of events depending on a finite number of increments.

Solution: It is clear that since the increments take value in $\hat{\mathbb{N}}$, that the Poisson process is piecewise constant, takes values in \mathbb{N} , it is non-decreasing and piecewise constant increasing only by jumps.

We show that jumps have size 1.

Consider the dyadics, $D_n = \{k2^{-n}, k \in \mathbb{N}\}$, $D = \bigcup_{n \in \mathbb{N}} D_n$.
Let

$$A_k^n = \{\omega : (N_{k2^{-n}} - N_{(k-1)2^{-n}}) \leq 1\}$$

$$P(A_k^n) = \exp(-\lambda 2^{-n}) \left(1 + \lambda 2^{-n}\right)$$

Let also

$$A^n = \bigcap_{k=1}^{2^n} A_k^n$$

By the independence of the increments

$$\begin{aligned} P(A^n) &= \prod_{k=1}^{2^n} P(A_k^n) = \exp(-\lambda 2^{-n})^{2^n} \left(1 + \lambda 2^{-n}\right)^{2^n} \\ &= \exp(-\lambda) \left(1 + \lambda 2^{-n}\right)^{2^n} \uparrow \exp(-\lambda) \exp(\lambda) = 1 \end{aligned}$$

as $n \rightarrow \infty$.

Moreover there is a subsequence n_m such that

$$0 \leq (1 - P(A_{n_m})) \leq 2^{-m}$$

which implies

$$\sum_{m=0}^{\infty} P((A_{n_m})^c) < \infty$$

by Borel Cantelli lemma

$$P(\liminf_m A^{n_m}) = 1 - P(\limsup_m (A^{n_m})^c) = 1$$

which means that P a.s., for all m large enough, for all $k = 1, \dots, 2^{n_m}$

$$(N_{k2^{-n_m}} - N_{(k-1)2^{-n_m}}) \leq 1$$

But since increments are non-negative this means P -almost surely and $\forall d \in D_n \cap [0, 1]$ with n large enough

$$(N_{d+2^{-n}} - N_d) \leq 1$$

But this implies that P -almost surely, for all $t \in [0, 1]$

$$0 \leq N_{d+}(\omega) - N_{d-}(\omega) \leq 1$$

where

$$N_{t+}(\omega) = \lim_{d \downarrow t, t \in D} N_d(\omega), \quad N_{t-}(\omega) = \lim_{d \uparrow t, t \in D} N_d(\omega)$$

- Compute the probability density of the first jump time $\tau(\omega)$. $P(\tau > t) = P(N_t = 0) = \exp(-\lambda t)$

$$p_\tau(t) = -\frac{d}{dt}P(\tau > t) = \exp(-\lambda t)\lambda$$

4. For a fixed $t \in [0, 1]$ consider the function $s \mapsto h_t(s) := (s \wedge t) = \min(s, t)$

- Show that $h_t(\cdot) = (t \wedge \cdot)$ belongs to the Cameron-Martin space H .

Solution

$$t \wedge s = \int_0^s \mathbf{1}(u \leq t) du$$

where $u \mapsto \mathbf{1}(u \leq t)$ is square integrable because it is bounded.

- Show that $B_t(\omega) := B(h_t)$ is a Brownian motion.
- Show that $K(t, s) := \text{Cov}(B_t, B_s) = E(B_t B_s) = s \wedge t$.

The covariance of

$$E(B_t B_s) = (h_t, h_s)_H = \int_0^1 \mathbf{1}_{[0,t]}(u) \mathbf{1}_{[0,s]}(u) ds = t \wedge s$$

and

$$E((B_t - B_s)(B_v - B_u)) = t \wedge v + u \wedge s - t \wedge u - s \wedge v$$

in case $0 \leq s \leq t \leq u \leq v$ we get $(t + s - t - s) = 0$, so disjoint increments are uncorrelated. In case $t = v$ $s = v$ we get $(t - s)$.

By definition of isonormal gaussian process the variables $(B_{t_1}, \dots, B_{t_d}) = (B(h_{t_1}), \dots, B(h_{t_d}))$ are jointly gaussian with zero mean and so are the increments, $(\Delta B_i = (B_{t_i} - B_{t_{i-1}}), i = 1, \dots, d)$, since they are obtained by a linear tranformation.

Therefore uncorrelated jointly gaussian increments are independent.

By isometry we obtain the series expansion with respect to the Haar basis

$$\begin{aligned} B_t &= \sum_{n \in \mathbb{N}} \sum_{d \in D_n \setminus D_{n-1}} B(\eta_d)(\eta_d, (\cdot \wedge t))_H = \\ &= \sum_{n \in \mathbb{N}} \sum_{d \in D_n \setminus D_{n-1}} B(\eta_d)(\dot{\eta}_d, \mathbf{1}_{[0,t]})_H = \\ &= \sum_{n \in \mathbb{N}} \sum_{d \in D_n \setminus D_{n-1}} B(\eta_d) \int_0^t \dot{\eta}_d(s) ds \\ &= \sum_{n \in \mathbb{N}} \sum_{d \in D_n \setminus D_{n-1}} B(\eta_d) \eta_d(t) \end{aligned}$$

where the series converges in $L^2(\Omega)$. In the Lévy construction we have shown also that the expansion converges P almost surely uniformly in $[0, 1]$, which implies P -a.s.continuity.

- Show the reproducing kernel Hilbert space property in the Cameron Martin space H : for $h \in H$

$$(K(t, \cdot), h(\cdot))_H = h(t)$$

Sol. We have seen that $K(t, s) = E(B_t B_s) = (t \wedge s)$

$$\begin{aligned} (K(t, \cdot), h(\cdot))_H &= ((t \wedge \cdot), h)_H = \\ (\mathbf{1}_{[0,t]}, \dot{h})_{L^2([0,1])} &= \int_0^t \dot{h}(s) ds = h(t) \end{aligned}$$