

Solutions to the exercises of Lecture 7, ISem 2015/2016

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Exercise (7.1). Let X be a separable Banach space endowed with a centred Gaussian measure γ . Prove that for any choice $h_1, \dots, h_d \in H$, the map $P: X \rightarrow \mathbb{R}^d$, $P(x) := (\hat{h}_1(x), \dots, \hat{h}_d(x))$ is a Gaussian random variable with law $P_{\#}(\gamma) = \mathcal{N}(0, Q)$, where $Q_{k,j} = [h_k, h_j]_H$.

Proof. First of all, notice that

$$\widehat{P_{\#}\gamma}(\xi) = \int_{\mathbb{R}^d} e^{ix\xi} (P_{\#}\gamma)(dx) = \int_X e^{iP(x)\cdot\xi} \gamma(dx) = \widehat{\gamma}(g),$$

where we put $g = P \cdot \xi = \sum_{k=1}^d \hat{h}_k \xi_k \in X_{\gamma}^*$. Next, observe that

$$\begin{aligned} \|g\|_{L^2(X,\gamma)}^2 &= \left\langle \sum_{k=1}^d \hat{h}_k \xi_k, \sum_{j=1}^d \hat{h}_j \xi_j \right\rangle_{L^2(X,\gamma)} \\ &= \sum_{j,k=1}^d \xi_k \xi_j \left\langle \hat{h}_k, \hat{h}_j \right\rangle_{L^2(X,\gamma)} = \sum_{j,k=1}^d \xi_k \xi_j [h_k, h_j]_H, \end{aligned}$$

since the map $h \mapsto \hat{h}$ is an isometry of H into $L^2(X, \gamma)$ by Proposition 3.1.2. Now, by Proposition 2.3.5,

$$\widehat{\gamma}(g) = e^{-\frac{1}{2}\|g\|_{L^2(X,\gamma)}^2} = e^{-\frac{1}{2}\sum_{j,k=1}^d \xi_k \xi_j [h_k, h_j]_H} = e^{-\frac{1}{2}\langle Q\xi, \xi \rangle},$$

where $Q_{k,j} = [h_k, h_j]_H$. □

Exercise (7.2, Part 1). Let $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}$ be a convex function. Then there exist a sequence (a_n) in \mathbb{R}^d and a sequence (b_n) in \mathbb{R} such that

$$\varphi(x) = \sup_{n \in \mathbb{N}} (a_n \cdot x + b_n).$$

Proof. We will need the following two well-known results (see *Rockafellar, Convex Analysis*, Theorems 10.1 and 23.4).

1. If $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}$ is a convex function, then φ is continuous.
2. If $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}$ is a convex function, then for every $y \in \mathbb{R}^d$ there exist $a_y \in \mathbb{R}^d$ and $b_y \in \mathbb{R}$ such that

$$\varphi(x) \geq a_y \cdot x + b_y$$

for every $x \in \mathbb{R}^d$, with equality when $x = y$. In other words, the subgradient of φ is non-empty at every point.

For every $y \in \mathbb{R}^d$ take $a_y \in \mathbb{R}^d$ and $b_y \in \mathbb{R}$ such that

$$\varphi(x) \geq a_y \cdot x + b_y$$

for every $x \in \mathbb{R}^d$, with equality when $x = y$. Let then (x_n) be any enumeration of \mathbb{Q}^d , and call $a_n := a_{x_n}$, $b_n := b_{x_n}$. Therefore,

$$\varphi(x) \geq a_n \cdot x + b_n$$

for every $x \in \mathbb{R}^d$ and every $n \in \mathbb{N}$. Hence,

$$\varphi(x) \geq \sup_n (a_n \cdot x + b_n) =: \psi(x)$$

for every $x \in \mathbb{R}^d$, with equality on \mathbb{Q}^d by construction. Now, both φ and ψ are convex, hence continuous, functions and they agree on a dense subset of \mathbb{R}^d ; therefore they coincide on the whole of \mathbb{R}^d . \square

Exercise (7.2, Part 2 - Jensen). Let $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}$ be a convex function, (Ω, X, \mathbb{P}) a probability space, \mathcal{G} a sub- σ -algebra of \mathcal{F} , and $X \in L^1(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$; assume that $\varphi \circ X \in L^1(\Omega, \mathcal{F}, \mathbb{P})$. Then

$$\varphi \circ \mathbb{E}(X|\mathcal{G}) \leq \mathbb{E}(\varphi \circ X|\mathcal{G}).$$

Proof. Take a sequence (a_n) in \mathbb{R}^d and a sequence (b_n) in \mathbb{R} such that

$$\varphi(x) = \sup_{n \in \mathbb{N}} (a_n \cdot x + b_n)$$

for every $x \in \mathbb{R}^d$. By monotonicity and linearity, one has $\mathbb{E}(\varphi \circ X|\mathcal{G}) \geq a_n \cdot \mathbb{E}(X|\mathcal{G}) + b_n$, so that, by taking the least upper bound,

$$\varphi \circ \mathbb{E}(X|\mathcal{G}) = \sup_{n \in \mathbb{N}} (a_n \cdot \mathbb{E}(X|\mathcal{G}) + b_n) \leq \mathbb{E}(\varphi \circ X|\mathcal{G}). \quad \square$$

Exercise (7.3). Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, \mathcal{G} a sub- σ -algebra of \mathcal{F} and $X, Y \in L^1(\Omega, \mathcal{F}, \mathbb{P})$. Then the following hold:

1. if $\mathcal{G} = \{\emptyset, \Omega\}$, then $\mathbb{E}(X|\mathcal{G}) = \mathbb{E}[X]$;
2. $\mathbb{E}[\mathbb{E}(X|\mathcal{G})] = \mathbb{E}[X]$;
3. if $X \leq Y$, then $\mathbb{E}(X|\mathcal{G}) \leq \mathbb{E}(Y|\mathcal{G})$;

4. for all $\alpha, \beta \in \mathbb{R}$, $\mathbb{E}(\alpha X + \beta Y | \mathcal{G}) = \alpha \mathbb{E}(X | \mathcal{G}) + \beta \mathbb{E}(Y | \mathcal{G})$;
5. if \mathcal{H} is a sub- σ -algebra of \mathcal{G} , then $\mathbb{E}(\mathbb{E}(X | \mathcal{G}) | \mathcal{H}) = \mathbb{E}(X | \mathcal{H})$;
6. if X is \mathcal{G} -measurable, then $\mathbb{E}(X | \mathcal{G}) = X$;
7. if $XY \in L^1(\Omega, \mathcal{F}, \mathbb{P})$ and X is \mathcal{G} -measurable, then $\mathbb{E}(XY | \mathcal{G}) = X \mathbb{E}(Y | \mathcal{G})$;
8. if X is independent of \mathcal{G} , then $\mathbb{E}(X | \mathcal{G}) = \mathbb{E}[X]$.

Proof. 1. We just observe that $\mathbb{E}[X]$ belongs to $L^1(\Omega, \mathcal{G}, \mathbb{P})$ and trivially satisfies the equality $\mathbb{E}[\mathbb{E}[X] \chi_B] = \mathbb{E}[X \chi_B]$ for every $B \in \{\emptyset, \Omega\}$.

2. Set $Y = \mathbb{E}(X | \mathcal{G})$. Then, the definition of conditional expectation implies that $\mathbb{E}[X \chi_B] = \mathbb{E}[Y \chi_B]$ for every \mathcal{G} -measurable set B . Take $B = \Omega$ to get the result.
3. Choose two representatives f, g of $\mathbb{E}(X | \mathcal{G}), \mathbb{E}(Y | \mathcal{G})$ respectively, and let, for every $n \in \mathbb{N}^*$, $E_n := \{f \geq g + \frac{1}{n}\}$; then $E_n \in \mathcal{G}$. Moreover,

$$\mathbb{E}(g \chi_{E_n}) = \mathbb{E}(Y \chi_{E_n}) \geq \mathbb{E}(X \chi_{E_n}) = \mathbb{E}(f \chi_{E_n}) \geq \mathbb{E}(g \chi_{E_n}) + \frac{1}{n} \mathbb{P}(E_n),$$

so that $\mathbb{P}(E_n) = 0$. Then

$$\{f > g\} = \bigcup_{n \in \mathbb{N}^*} E_n$$

has \mathbb{P} -measure zero. Therefore $f \leq g$ \mathbb{P} -a.s., i.e. $\mathbb{E}(X | \mathcal{G}) \leq \mathbb{E}(Y | \mathcal{G})$.

4. First notice that $\alpha \mathbb{E}(X | \mathcal{G}) + \beta \mathbb{E}(Y | \mathcal{G})$ is \mathcal{G} -measurable. Moreover, for every $B \in \mathcal{G}$

$$\begin{aligned} \mathbb{E}[(\alpha X + \beta Y) \chi_B] &= \alpha \mathbb{E}[X \chi_B] + \beta \mathbb{E}[Y \chi_B] \\ &= \alpha \mathbb{E}[\mathbb{E}(X | \mathcal{G}) \chi_B] + \beta \mathbb{E}[\mathbb{E}(Y | \mathcal{G}) \chi_B] \\ &= \mathbb{E}[(\alpha \mathbb{E}(X | \mathcal{G}) + \beta \mathbb{E}(Y | \mathcal{G})) \chi_B], \end{aligned}$$

whence the result.

5. First notice that $\mathbb{E}(\mathbb{E}(X | \mathcal{G}) | \mathcal{H})$ is \mathcal{H} -measurable. Moreover, if $B \in \mathcal{H}$ then $B \in \mathcal{G}$, so that $\mathbb{E}[X \chi_B] = \mathbb{E}[\mathbb{E}(X | \mathcal{G}) \chi_B] = \mathbb{E}[\mathbb{E}(\mathbb{E}(X | \mathcal{G}) | \mathcal{H}) \chi_B]$, whence the result.
6. X itself satisfies the properties of the definition of $\mathbb{E}(X | \mathcal{G})$.

7. First notice that $X\mathbb{E}(Y|\mathcal{G})$ is \mathcal{G} -measurable. Observe also that by Remark 7.2.2 the statement holds for every \mathcal{G} -measurable simple function. Let (X_n) be a sequence of \mathcal{G} -measurable simple functions which converges \mathbb{P} -almost everywhere to X such that $|X_n| \leq |X|$ \mathbb{P} -almost everywhere. By the dominated convergence theorem, for every $B \in \mathcal{G}$

$$\mathbb{E}[XY\chi_B] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n Y \chi_B] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n \mathbb{E}(Y|\mathcal{G})\chi_B] = \mathbb{E}[X\mathbb{E}(Y|\mathcal{G})\chi_B],$$

whence the result.

8. Assume first that X is the characteristic function of some $E \in \mathcal{F}$. Then, for every $B \in \mathcal{G}$,

$$\begin{aligned} \mathbb{E}[X\chi_B] &= \mathbb{E}[\chi_E\chi_B] = \mathbb{P}(E \cap B) \\ &= \mathbb{P}(E)\mathbb{P}(B) = \mathbb{E}[\mathbb{E}[\chi_E]\chi_B] = \mathbb{E}[\mathbb{E}[X]\chi_B]. \end{aligned}$$

Thus $\mathbb{E}(X|\mathcal{G}) = \mathbb{E}[X]$. By linearity the statement holds for any \mathcal{F} -measurable simple function which is a finite linear combination of characteristic functions of sets independent of \mathcal{G} . Now, let X be independent of \mathcal{G} , and let (X_n) be a sequence of $\mathcal{E}(\Omega, X)$ -measurable simple functions which converges pointwise a.e. to X such that $|X_n| \leq |X|$ \mathbb{P} -a.e. for every $n \in \mathbb{N}$. Then, for every $B \in \mathcal{G}$, the dominated convergence theorem implies that

$$\mathbb{E}[X\chi_B] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n\chi_B] = \lim_{n \rightarrow \infty} \mathbb{E}[X_n]\mathbb{E}[\chi_B] = \mathbb{E}[\mathbb{E}[X]\chi_B]. \quad \square$$

Exercise (7.4). Prove that, if $\Omega = (0, 1)^2$, $\mathcal{F} = \mathcal{B}((0, 1)^2)$, $\mathbb{P} = \lambda_2$ is the Lebesgue measure on Ω , $X \in L^1(\Omega, \mathcal{F}, \mathbb{P})$, and $\mathcal{G} = \mathcal{B}((0, 1)) \times (0, 1)$, then

$$\mathbb{E}(X|\mathcal{G})(x, y) = \int_0^1 X(x, t) d\lambda_1(t) \quad \text{for a.e. } (x, y) \in \Omega.$$

Proof. Since $\mathcal{B}((0, 1)^2) = \mathcal{B}(0, 1) \otimes \mathcal{B}(0, 1)$, by Fubini's theorem the function

$$Y = \int_0^1 X(\cdot, t) d\lambda_1(t) \otimes 1$$

belongs to $L^1(\Omega, \mathcal{G}, \mathbb{P})$.

Let $A \in \mathcal{G}$; then A has the form $A = A_1 \times (0, 1)$ for some $A_1 \in \mathcal{B}(0, 1)$, so that $\chi_A(x, y) = \chi_{A_1}(x)$ for every $y \in (0, 1)$. Observe also that $\lambda_2 = \lambda_1 \otimes \lambda_1$.

Therefore,

$$\begin{aligned}
\mathbb{E}(Y\chi_A) &= \int_0^1 \int_0^1 \int_0^1 X(x,t)d\lambda_1(t)\chi_A(x,y)d\lambda_1(x)d\lambda_1(y) \\
&= \int_0^1 \int_0^1 X(x,t)d\lambda_1(t)\chi_{A_1}(x)d\lambda_1(x) \\
&= \int_0^1 \int_0^1 X(x,y)\chi_{A_1}(x)d\lambda_1(x)d\lambda_1(y) \\
&= \int_0^1 \int_0^1 X(x,y)\chi_A(x,y)d\lambda_1(x)d\lambda_1(y) = \mathbb{E}(X\chi_A).
\end{aligned}$$

The proof is then complete. \square

Exercise (7.5). Prove that, for every fixed sequence $(f_n) \subset X^*$, the family of sets

$$\mathcal{A} = \{E = \{x \in X : (f_n(x)) \in B\} : B \in \mathcal{B}(\mathbb{R}^\infty)\}$$

is a σ -algebra.

Proof. First of all $\emptyset \in \mathcal{A}$, since $\emptyset = \{x \in X : (f_n(x)) \in \emptyset\}$.

Let $E = \{x \in X : (f_n(x)) \in B\}$ for some $B \in \mathcal{B}(\mathbb{R}^\infty)$. Then

$$E^c = \{x \in X : (f_n(x)) \notin B\} = \{x \in X : (f_n(x)) \in B^c\} \in \mathcal{A}.$$

If $\{E_j\} \subset \mathcal{A}$ and $E_j = \{x \in X : (f_n(x)) \in B_j\}$, then

$$\bigcup_j E_j = \{x \in X : \exists j : (f_n(x)) \in B_j\} = \{x \in X : (f_n(x)) \in \cup_j B_j\} \in \mathcal{A}.$$

This completes the proof. \square

Exercise (7.6). Prove that if X is an infinite dimensional Banach space and $\varphi \in C^\infty(X)$ has compact support, then $\varphi \equiv 0$.

Proof. Assume that there is some $x \in X$ such that $\varphi(x) \neq 0$. By continuity, there exists a closed ball \bar{B} such that $\varphi(x) \neq 0$ for every $x \in \bar{B}$. Thus \bar{B} is contained in the support of φ . Since the latter is compact, and \bar{B} is closed, \bar{B} is compact. Therefore X is finite dimensional. \square

Remark. In Exercise 7.6 differentiability plays no role, and the same result holds for $\varphi \in C_c(X)$.