Institute for Applied Mathematics Winter term 2016/17

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"Markov Processes", Problem Sheet 1.

Hand in solutions before Monday 24.10., 2 pm. (post-box opposite to maths library)

Tutorial classes on wednesdays, 8-10 and 12-14, in 1.008. First class: 19.10.

1. (Conditional Expectations 1)

Let X, Y on (Ω, \mathcal{A}, P) be independent random variables that are Bernoulli distributed with parameter p. We set $Z = 1_{\{X+Y=0\}}$. Compute E(X|Z) and E(Y|Z). Are these random variables still independent?

2. (Conditional Expectations 2)

a) Let X, Y, Z be random variables with values in a measurable space (S, \mathcal{B}) such that the couples (X, Z) and (Y, Z) have the same law. Show that, for any $f \in \mathcal{F}_+(S)$,

$$E(f(X)|Z) = E(f(Y)|Z)$$
 a.s.

- b) Let T_1, \ldots, T_n be i.i.d. real integrable random variables. Set $T = T_1 + \cdots + T_n$.
 - i) Show that $E(T_1|T) = T/n$.
 - ii) Compute $E(T|T_1)$.

3. (Conditional laws of Gaussian vectors)

- a) Let X be a r.v. with values in \mathbb{R}^m of the form $X = \phi(Y) + Z$, where Y and Z are independent. Show that the conditional law of X given Y = y coincides with the law of $\phi(y) + Z$.
- b) Let X, Y be random variables taking values in \mathbb{R}^k and \mathbb{R}^p , respectively. We assume that their joint law in $(\mathbb{R}^{k+p}, \mathcal{B}(\mathbb{R}^{k+p}))$ is Gaussian with mean and covariance matrix given, respectively, by

$$\begin{pmatrix} m_X \\ m_Y \end{pmatrix}, \quad \begin{pmatrix} R_X & R_{XY} \\ R_{YX} & R_Y \end{pmatrix}.$$

Here m_X , m_Y , and R_X , R_Y are the means and covariance matrices of X and Y, respectively, and $R_{X,Y} = E((X - E(X))(Y - E(Y))^T) = R_{YX}^T$ is the $k \times p$ matrix of the covariances between components of X and Y. We assume that R_Y is strictly positive definite.

- i) Find a $k \times p$ matrix A such that the random variables X AY and Y are independent.
- (ii) Show that the conditional law of X given Y is Gaussian with mean $E(X|Y) = m_X + R_{XY}R_Y^{-1}(Y m_Y)$ and covariance matrix $R_X R_{XY}R_Y^{-1}R_{YX}$.
- c) Let X be a signal with normal law $\mathcal{N}(0,1)$. We assume that we cannot observe the value of X; instead we only observe the value of Y = X + W, where W is independent of X with law $\mathcal{N}(0, \sigma^2)$.
 - i) Given an observation Y = y, give an estimate of the value X of the signal.
 - ii) Let us assume that $\sigma^2 = 0.1$ and that the value of the observation is Y = 0.55. What is the probability for the signal X to be in the interval $\left[\frac{1}{4}, \frac{3}{4}\right]$?

4. (Martingales)

- a) Let $X = (X_n)_{n \in \mathbb{Z}_+}$ be a supermartingale such that $E(X_n)$ is constant. Show that X is a martingale.
- b) Let $X = (X_n)_{n \in \mathbb{Z}_+}$ be an integrable process adapted to the filtration (\mathcal{F}_n) . Show that X is an (\mathcal{F}_n) -martingale if and only if $E(X_T) = E(X_0)$ for any bounded (\mathcal{F}_n) stopping time T.

5. (Martingales with independent increments)

A process $(M_n)_{n\in\mathbb{Z}_+}$ is said to be with independent increments if, for every n, the r.v. $M_{n+1}-M_n$ is independent of the σ -algebra $\mathcal{F}_n=\sigma(M_0,\ldots,M_n)$.

- a) Let $(M_n)_{n\in\mathbb{Z}_+}$ be a square integrable martingale with independent increments. We set $\sigma_0^2 = \text{Var}(M_0)$ and, for $k \geq 1$, $\sigma_k^2 = \text{Var}(M_k M_{k-1})$.
 - i) Show that $Var(M_n) = \sum_{k=0}^n \sigma_k^2$.
 - ii) Determine the conditional variance process $\langle M \rangle_n$ of (M_n) .
- b) Let (M_n) be a Gaussian martingale (we recall that a process $(M_n)_{n\geq 0}$ is Gaussian if, for every n, the vector (M_0, \ldots, M_n) is Gaussian).
 - i) Show that $(M_n)_{n\geq 0}$ has independent increments.
 - ii) Show that, for every fixed $\theta \in \mathbb{R}$, the process $Z_n^{\theta} = \exp\left(\theta M_n \frac{1}{2}\theta^2 \langle M \rangle_n\right)$ is a martingale. Does it converge almost surely?