

**Malliavin calculus, exercises 1, 26.01.2011**

1. Check Abel summation formula: for  $(x_0, x_1, \dots, x_n), (y_0, y_1, \dots, y_n) \in \mathbb{R}^{n+1}$ , Denoting  $\Delta x_i = (x_i - x_{i-1}), \Delta y_i = (y_i - y_{i-1})$ , check Abel summation formula

$$x_n y_n - x_0 y_0 = \sum_{i=1}^n x_{i-1} \Delta y_i + \sum_{i=1}^n y_{i-1} \Delta x_i + \sum_{i=1}^n \Delta y_i \Delta x_i$$

2. • Use Fubini theorem to prove the integration by parts formula

$$F(b)G(b) - F(a)G(a) = \int_a^b G(x)F'(dx) + \int_a^b F(x)G'(dx)$$

when  $F$  and  $G$  have finite variation, and have no common jumps  $(F(x+) - F(x-))(G(x+) - G(x-)) = 0 \forall x$  (you can assume that  $F$  and  $G$  are continuous).

- What is the formula when there are common jumps ? (look at Abel discrete formula)

3. Let  $G$  a standard gaussian random variable  $E(G) = 0, E(G^2) = 1$ .

- Use integration by parts and induction to compute the moments  $E(G^n) \ n \in \mathbb{N}$ .
- Show that  $G$  and  $(G^2 - 1)$  are uncorrelated.
- Let  $X$  be  $\mathcal{G}$ -measurable and  $Y \stackrel{P}{\perp\!\!\!\perp} \mathcal{G}$ . Show that

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

- Check this property of the conditional expectation using Kolmogorov definition:

If the  $\sigma$ -algebra  $\mathcal{H}$  is  $P$ -independent w.r.t the  $\sigma$ -algebra  $\sigma(X) \vee \mathcal{G}$ , then

$$E_P(X | \sigma(G) \vee \sigma(H))(\omega) = E_P(X | \sigma(G))(\omega)$$

- Consider the operator  $\partial^*$  where  $\partial^* f(x) = f(x)x - f'(x)$ .  
Denote  $h_0(x) \equiv 1$ ,  
 $h_n(x) = \partial^* h_{n-1}(x) = (\partial^*)^n h_0(x) = (\partial^*)^n 1$ .  
Use integration by parts formula  $E(\partial f(G)h(G)) = E(f(G)\partial^* h(G))$   
Show that  $E(h_n(G)) = 0$  and  $E(h_n(G)h_m(G)) = 0$  when  $n \neq m$ .  
 $h_n(x)$  are (unnormalized) Hermite polynomials.

4. Consider random variables  $X(\omega), Y(\omega), Z(\omega) \in L^2(\Omega, \mathcal{F}, P)$ . Let

$$H = \{a(Y) + b(Y)Z : a, b \text{ Borel measurable functions}\} \cap L_2(\Omega, \sigma(Y, Z), P)$$

the subspace random variables in  $L^2(\Omega, \sigma(Z, Y), P)$  which depend linearly on  $Z$ .

- Show that

$$\hat{X}(\omega) = E_P(X|\sigma(Y)) + \left( Z(\omega) - E_P(Z|\sigma(Y))(\omega) \right) \frac{\text{Cov}_P(X, Z|\sigma(Y))(\omega)}{\text{Var}_P(Z|\sigma(Y))(\omega)}$$

is the orthogonal projection of  $X$  to the subspace  $H$ .

- Compute the conditional square error  $E_P((\hat{X} - X)^2|\sigma(Y))(\omega)$ .
- Compute the square error  $E_P((\hat{X} - X)^2)$ .

5. Let  $(X, Y)$  jointly gaussian,  $E(X) = m_X$ ,  $E(Y) = m_Y$  and covariance matrix

$$\Sigma = \begin{pmatrix} \Sigma_{XX} & \Sigma_{XY} \\ \Sigma_{XY} & \Sigma_{YY} \end{pmatrix}$$

- Use Bayes formula to compute the conditional distribution of  $X$  conditionally on  $\sigma(Y)$ .
- Show that the conditional expectation has the form

$$E(X|\sigma(Y))(\omega) = \hat{b} + \hat{a}Y(\omega)$$