UNIVERSITÀ DI ROMA TOR VERGATA

Laurea Magistrale in Matematica Pura e Applicata

Corso di *EP-Elementi di Probabilità* P.Baldi Esonero 1 del 22 aprile 2016

Exercise 1 Let $B = (\Omega, \mathcal{F}, (\mathcal{F}_t)_t, (B_t)_t, P)$ be a Brownian motion.

a) Prove that there exists $\alpha \in \mathbb{R}$ such that

$$X_t = B_t^4 - 6t B_t^2 + \alpha t^2$$

is a martingale. (Recall: $(x+y)^4 = x^4 + 4x^3y + 6x^2y^2 + 4xy^3 + y^4$, $E[Z^4] = 3$ if $Z \sim N(0, 1)$).

b) Let a > 0 and let τ the exit time of *B* from] - a, a[. Recall that we know already that τ is integrable and that $E[\tau] = a^2$. Compute $E[\tau^2]$ and $Var(\tau)$.

Exercise 2 Let $B = (\Omega, \mathcal{F}, (\mathcal{F}_t)_t, (B_t)_t, P)$ be a Brownian motion. Recall that, for every $\lambda \in \mathbb{R}$,

$$M_t = \mathrm{e}^{\lambda B_t - \frac{1}{2}\,\lambda^2 t}$$

is a positive martingale with expectation equal to 1. For T > 0 let us define on (Ω, \mathcal{F}) a probability Q by, for every $A \in \mathcal{F}$,

$$\mathbf{Q}(A) = \mathbf{E}[M_T \mathbf{1}_A] = \int_A M_T \, d\mathbf{P} \, .$$

Let us denote E^Q the expectation with respect to Q.

a) Prove that if $t \leq T$ and $A \in \mathcal{F}_t$

$$\mathbf{Q}(A) = \mathbf{E}[M_t \mathbf{1}_A]$$

b1) Prove that, for every $z \in \mathbb{C}$,

$$Z_t = \mathrm{e}^{zB_t - \frac{1}{2}z^2t}$$

is a complex martingale.

b2) Compute $E^{Q}[e^{i\theta B_{t}}]$, for $\theta \in \mathbb{R}$. Is B a Brownian motion also with respect to Q?

b3) Prove that, with respect to Q, $X_t = B_t - \lambda t$ is a Brownian motion for $t \in [0, T]$.

Exercise 3 Let $B = (\Omega, \mathcal{F}, (\mathcal{F}_t)_t, (B_t)_t, P)$ be a *m*dimensional Brownian motion and

$$X_t = B_{g(t)}$$

where $g(t) = 1 - e^{-t}$.

a) Prove that *X* is a Markov process and compute its transition function. Is it time homogeneous?

b) Let $(\Omega, \mathcal{F}, (\mathcal{F}_s)_s, (X_t)_t, (\mathbb{P}^{x,s})_{x,s})$ a realization of the Markov process associated to the transition function computed in a). Prove that, for every starting point *x* and initial time *s*, $(X_t)_t$ converges as $t \to +\infty$ to a limit distribution to be determined, possibly depending on *x*, *s*.

c) Compute, for every $f \in \mathscr{C}^2_K(\mathbb{R}^m)$, the generator $A_s f(x)$.

Solutions

Exercise 1. a) As the expectation of a martingale is constant, in order for *M* to be a martingale we must have

$$\mathbf{E}[B_t^4] + \alpha t^2 = 6t \mathbf{E}[B_t^2] \,.$$

As $B_t^4 \sim t^2 Z^4$ with $Z \sim N(0, 1)$, the previous relation becomes $3t^2 + \alpha t^2 = 6t^2$, hence $\alpha = 3$. Let us prove that with this choice of αX is actually a martingale. We have, for $s \leq t$,

$$E[B_t^4 | \mathcal{F}_s] = E[(B_s + (B_t - B_s))^4 \mathcal{F}_s] =$$

$$= B_s^4 + 4E[B_s^3(B_t - B_s)\mathcal{F}_s] + 6E[B_s^2(B_t - B_s)^2\mathcal{F}_s] + 4E[B_s(B_t - B_s)^3\mathcal{F}_s] + |E[(B_t - B_s)^4\mathcal{F}_s] =$$

$$= B_s^4 + 4B_s^3E[(B_t - B_s)\mathcal{F}_s] + 6B_s^2E[(B_t - B_s)^2\mathcal{F}_s] + 4B_sE[(B_t - B_s)^3\mathcal{F}_s] + |E[(B_t - B_s)^4\mathcal{F}_s] =$$

$$= B_s^4 + 4B_s^3E[(B_t - B_s)] + 6B_s^2E[(B_t - B_s)^2] + 4B_sE[(B_t - B_s)^3] + |E[(B_t - B_s)^4] =$$

$$= B_s^4 + 6B_s^2(t - s) + 3(t - s)^2$$

and also

$$E[tB_t^2\mathcal{F}_s] = tE[(B_s + (B_t - B_s)^2\mathcal{F}_s] = t(B_s^2 + 2E[B_s(B_t - B_s)\mathcal{F}_s] + E[(B_t - B_s)^2\mathcal{F}_s]) = t(B_s^2 + 2B_sE[(B_t - B_s)\mathcal{F}_s] + E[(B_t - B_s)^2]) = t(B_s^2 + (t - s)).$$

Therefore

$$E[X_t | \mathcal{F}_s] = B_s^4 + 6B_s^2(t-s) + 3(t-s)^2 - 6t(B_s^2 + (t-s)) + 3t^2 = B_s^4 - 6sB_s^2 + 3s^2 = X_s.$$

b) For every t > 0 we have, thanks to the stopping theorem,

(1)
$$0 = \mathbb{E}[X_{t \wedge \tau}] = \mathbb{E}[B_{t \wedge \tau}^4] - 6\mathbb{E}[(t \wedge \tau)B_{t \wedge \tau}^2] + 3\mathbb{E}[(t \wedge \tau)^2]$$

We have

$$\lim_{t \to +\infty} \mathbb{E}[(t \wedge \tau)^2] = \mathbb{E}[\tau^2]$$

by Beppo Levi's theorem. Also, by Lebesgue's theorem,

$$\lim_{t \to +\infty} \mathbb{E}[B_{t \wedge \tau}^4] = \mathbb{E}[B_{\tau}^4] = a^4$$
$$\lim_{t \to +\infty} \mathbb{E}[(t \wedge \tau)B_{t \wedge \tau}^2] = \mathbb{E}[\tau B_{\tau}^2] = a^2 \mathbb{E}[\tau] = a^4$$

Actually we have $|B_{t\wedge\tau}| \le a$, $(t\wedge\tau)B_{t\wedge\tau}^2 \le \tau a$ and we know already that τ is integrable. Therefore, taking the limit as $t \to +\infty$ (1) gives

$$0 = a^4 - 6a^4 + 3E[\tau^2]$$

i.e.

$$\mathrm{E}[\tau^2] = \frac{5}{3} a^4$$

and $\operatorname{Var}(\tau) = \operatorname{E}[\tau^2] - \operatorname{E}[\tau]^2 = \frac{5}{3}a^4 - a^4 = \frac{2}{3}a^4.$

Exercise 2. a) We have

$$Q(A) = E[M_T 1_A] = E[E[M_T 1_A | \mathcal{F}_t]] = E[1_A E[M_T | \mathcal{F}_t]] = E[M_t 1_A]$$

b) We have

$$E^{Q}[e^{i\theta B_{t}}] = E[M_{t}e^{i\theta B_{t}}] = e^{-\frac{1}{2}\lambda^{2}t}E\left[e^{(\lambda+i\theta)B_{t}}\right] = e^{-\frac{1}{2}(\lambda^{2}-(\lambda+i\theta)^{2})t} = e^{-\frac{1}{2}t\theta^{2}}e^{-i\lambda\theta t}$$

We recognize the characteristic function of a $N(\lambda t, t)$ distributed r.v. Therefore *B* is not a Brownian motion with respect to Q (B_t is not centered).

c1) Very much similar to the case $z \in \mathbb{R}$ that we have seen in class: we have, for $s \leq t$,

$$E[Z_t | \mathcal{F}_s] = e^{-\frac{1}{2}z^2t} E[e^{z(B_s + (B_t - B_s))} | \mathcal{F}_s] = e^{zB_s - \frac{1}{2}z^2t} E[e^{z(B_t - B_s)} | \mathcal{F}_s] = e^{zB_s - \frac{1}{2}z^2t} E[e^{z(B_t - B_s)}] = e^{zB_s - \frac{1}{2}z^2t + \frac{1}{2}z^2(t-s)} = Z_s.$$

c2) The simplest way of checking that $X_t = B_t - \lambda t$ is a Brownian motion is (recall Theorem 4.34) to verify that

$$Y_t = e^{i\theta X_t + \frac{1}{2}\theta^2 t} = e^{i\theta (B_t - \lambda t) + \frac{1}{2}\theta^2 t}$$

is a Q-martingale for every $\theta \in \mathbb{R}$. If $s \le t \le T$ and $A \in \mathcal{F}_s$, we have, again using a) and c1) for $z = \lambda + i\theta$,

$$\mathbf{E}^{\mathbf{Q}}[Y_t \mathbf{1}_A] = \mathbf{E}[Y_t \mathbf{1}_A M_t] =$$

$$= \mathbf{E}[\mathbf{e}^{i\theta(B_t - \lambda t) + \frac{1}{2}\theta^2 t} \mathbf{e}^{\lambda B_t - \frac{1}{2}\lambda^2 t} \mathbf{1}_A] = \mathbf{E}[\mathbf{e}^{(\lambda + i\theta)B_t - \frac{1}{2}t(\lambda + i\theta)^2} \mathbf{1}_A] =$$

$$= \mathbf{E}[\mathbf{e}^{(\lambda + i\theta)B_s - \frac{1}{2}s(\lambda + i\theta)^2} \mathbf{1}_A] = \mathbf{E}[Y_s M_s \mathbf{1}_A] = \mathbf{E}[Y_s M_T \mathbf{1}_A] = \mathbf{E}^{\mathbf{Q}}[Y_s \mathbf{1}_A].$$

Exercise 3. a) There are (at least) two possible ways of reasoning. First remark that X is clearly a Gaussian process for which we know a Markovianity criterion. Let us compute first the covariance function of X: as g is an increasing function, for $s \le t$,

$$K_{s,t} = \operatorname{Cov}(X_s, X_t) = \operatorname{E}[B_{g(s)}B_g(t)] = g(s) \wedge g(t) = g(s)$$

Therefore we have, for $u \leq s \leq t$,

$$K_{u,s}K_{s,s}^{-1}K_{s,t} = g(u)\frac{1}{g(s)}g(s) = g(u) = K_{u,t}$$

which ensures that X is a Markov process with respect to its natural filtration. In order to determine the transition function p(s, t, x, dy), let us recall that this is simply the conditional

distribution of X_t given $X_s = x$. With the well-known formulas of the conditional distributions of jointly Gaussian r.v.'s p(s, t, x, dy) is Gaussian with mean

$$\operatorname{E}[X_t] + \frac{K_{s,t}}{K_{s,s}} \left(x - \operatorname{E}[X_s] \right) = x$$

and variance

$$K_{t,t} - \frac{K_{s,t}^2}{K_{s,s}} = g(t) - g(s) = e^{-s} - e^{-t}$$

i.e. $p(s, t, x, dy) \sim N(x, e^{-s} - e^{-t}).$

Second method. We can here check directly the Markov property with respect to the filtration $(\mathcal{F}_{g(t)})_t$. Actually, thanks to the freezing lemma and with the trick of decomposing into sum of the actual position and of the increment, we have for every bounded measurable function f

(2)
$$E[f(B_{g(t)}) | \mathcal{F}_{g(s)}] = E[f(B_{g(t)} - B_{g(s)} + B_{g(s)}) | \mathcal{F}_{g(s)}] = \Phi(B_{g(s)})$$

where

$$\Phi(x) = \mathbf{E}\Big[f(\underbrace{B_{g(t)} - B_{g(s)} + x})\Big].$$

Therefore (2) proves that X is a Markov process with respect to its natural filtration $(\mathcal{F}_{g(t)})_t$ and, remarking that the r.v. indicated by the down brace is N(x, g(t) - g(s))-distributed, this proves also that this is the transitions function.

b) The law of X_t starting at (s, x) is $p(s, t, x, dy) \sim N(x, e^{-s} - e^{-t})$, which, as $t \to +\infty$, converges to a $N(x, e^{-s})$ distribution.

c) If $f \in \mathscr{C}^2_K(\mathbb{R}^m)$, the value of $A_s f(x)$ is given by

$$\lim_{h\to 0}\frac{1}{h}\left(T_{s,s+h}f(x)-f(x)\right)$$

(provided the limit exists) where

$$T_{s,s+h}f(x) = \int_{\mathbb{R}^m} f(y) p(s,s+h,x,dy) \, dx$$

The crucial remark is that p(s, s + h, x, dy) is the distribution of $x + B_{g(s+h)} - B_{g(s)} \sim x + B_{g(s+h)-g(s)}$, so that

$$A_{s}f(x) = \lim_{h \to 0} \frac{1}{h} \left(\mathbb{E}[f(x + B_{g(s+h)-g(s)}) - f(x)] \right) =$$

=
$$\lim_{h \to 0} \frac{1}{h} \frac{g(s+h) - g(s)}{g(s+h) - g(s)} \left(\mathbb{E}[f(x + B_{g(s+h)-g(s)}) - f(x)] \right).$$

As

$$\lim_{h \to 0} \frac{g(s+h) - g(s)}{h} = \lim_{h \to 0} \frac{1 - e^{-(s+h)} - 1 + e^{-s}}{h} = e^{-s} \lim_{h \to 0} \frac{1 - e^{-h}}{h} = e^{-s}$$

we have, going back to the generator of the Brownian motion that we already know,

$$A_{s}f(x) = e^{-s} \lim_{h \to 0} \frac{1}{g(s+h) - g(s)} \left(\mathbb{E}[f(x+B_{g(s+h) - g(s)}) - f(x)] \right) =$$

= $e^{-s} \lim_{t \to 0} \frac{1}{t} \left(\mathbb{E}[f(x+B_{t}) - f(x)] \right) = \frac{e^{-s}}{2} \Delta f(x) .$